

Multivariate analysis and complex networks

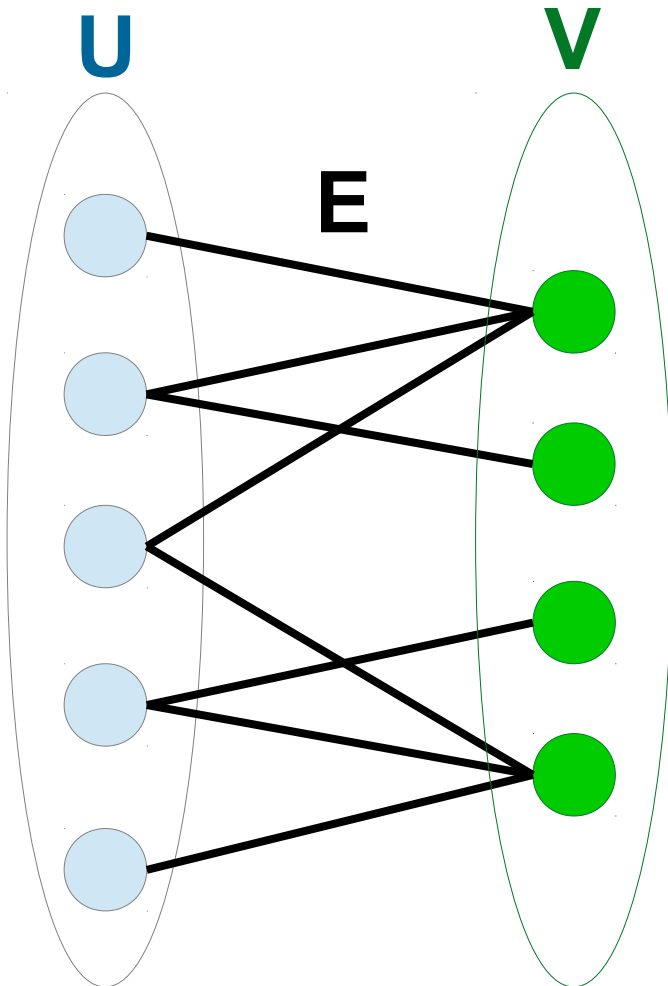
2: Bipartite Networks & Projections

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Bipartite Graphs



Bipartite graph:

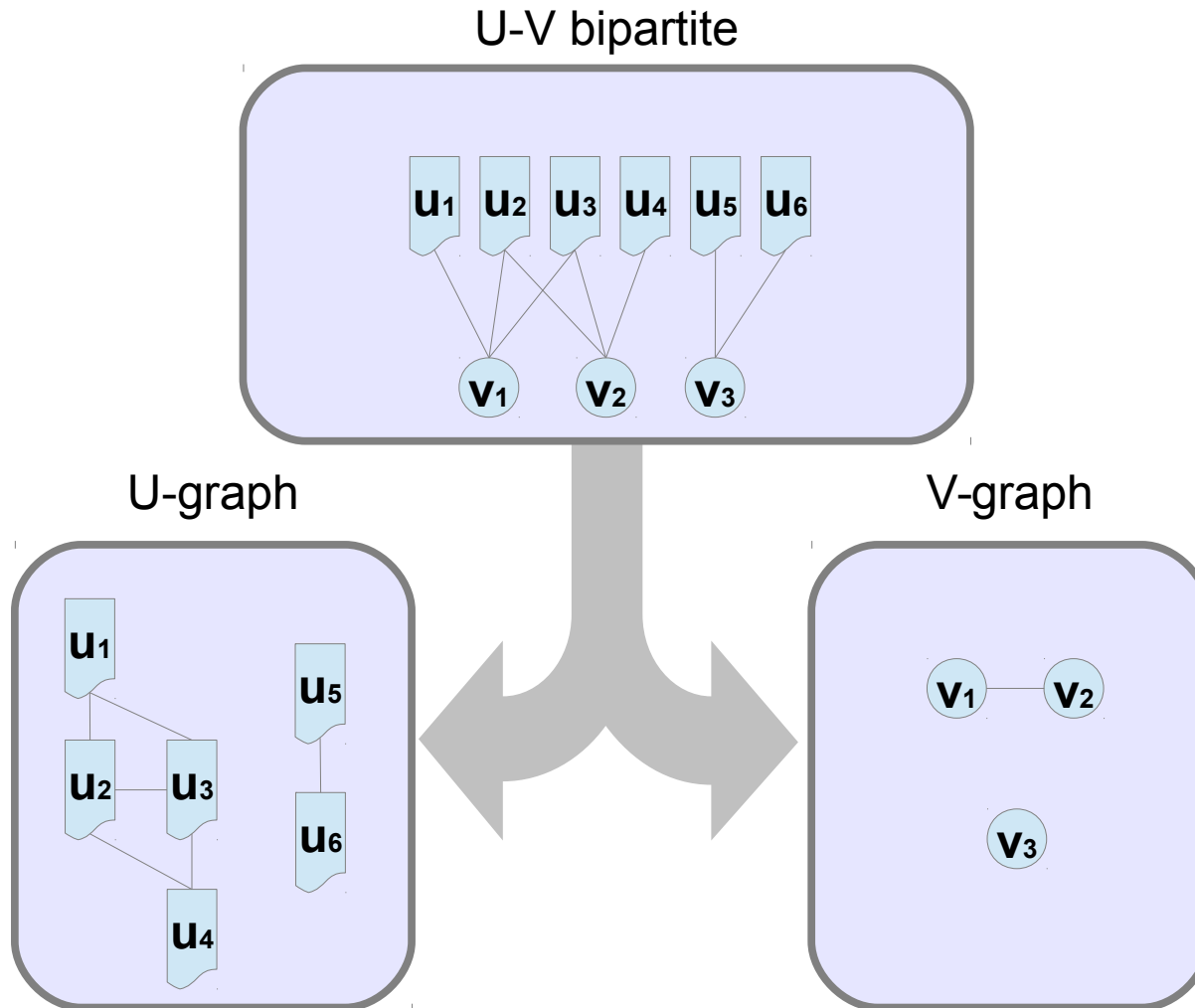
- $G = (U, V, E)$
- U, V nodes
- E edges among U, V

Bipartite graphs arise naturally when modelling relations between two different classes of entities

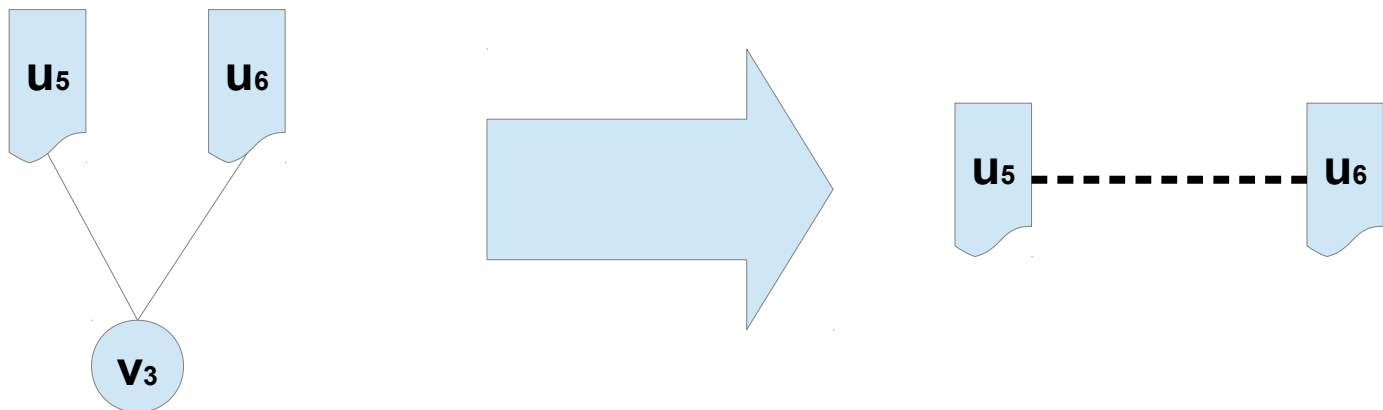
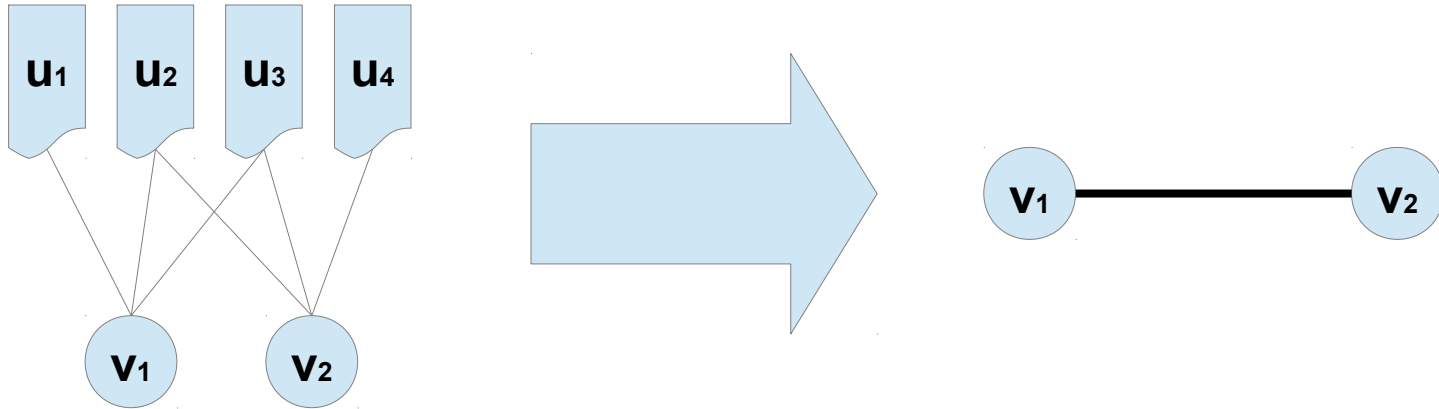
Examples

- Film – Actors
- Papers – Authors
- Nation – Products
- Text – Words
- Birds – Islands
- Plants – Pollinators
- Politician – Votes
- Users – Newsfeeds
- Patients – Symptoms
- Proteins - xxxx

Projections



Projections

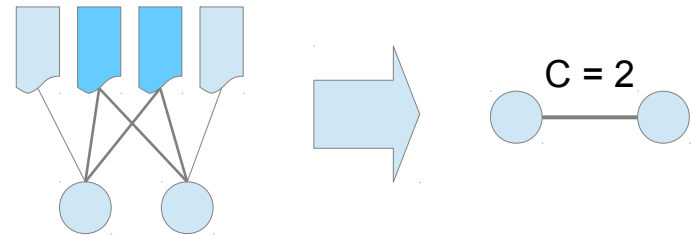


Co-occurrence matrix

- B = adjacency matrix of bipartite graph $G = (U, V, E)$
- $B_{uv} = 1$ if v has feature u , $B_{uv} = 0$ otherwise

The co-occurrence C_{uw} counts the number of times two features u, w occur together

$$C_{uw} = \sum_v B_{uv} B_{wv}$$



- $B B^T \rightarrow$ weighted adjacency matrix of the projection graph on U
- $B^T B \rightarrow$ weighted adjacency matrix of the projection graph on V

Null model

- C_{uw} = numbers of common neighbors of u, w
- n = maximum possible number of links
- $d_u = C_{uu}$ = degree of u
- $f_u = d_u / n$ fraction of possible links present
- If nodes were chosen at random:

$$f_{uv} = C_{uw} / n \rightarrow f_u f_v$$

$$P_{uw}(C) = \binom{C}{n} (f_u f_v)^C (1 - f_u f_v)^{n-C}$$

Other Projections

- Similarity Matrix

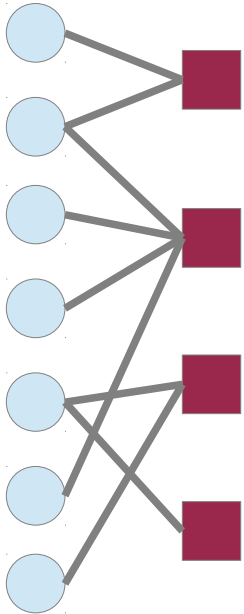
$$S_{uw} = 2 C_{uw} / (C_{uu} + C_{ww})$$

- Correlation matrix

$$\rho_{uv} = (f_{uv} - f_u f_v) / \sigma_u \sigma_v$$

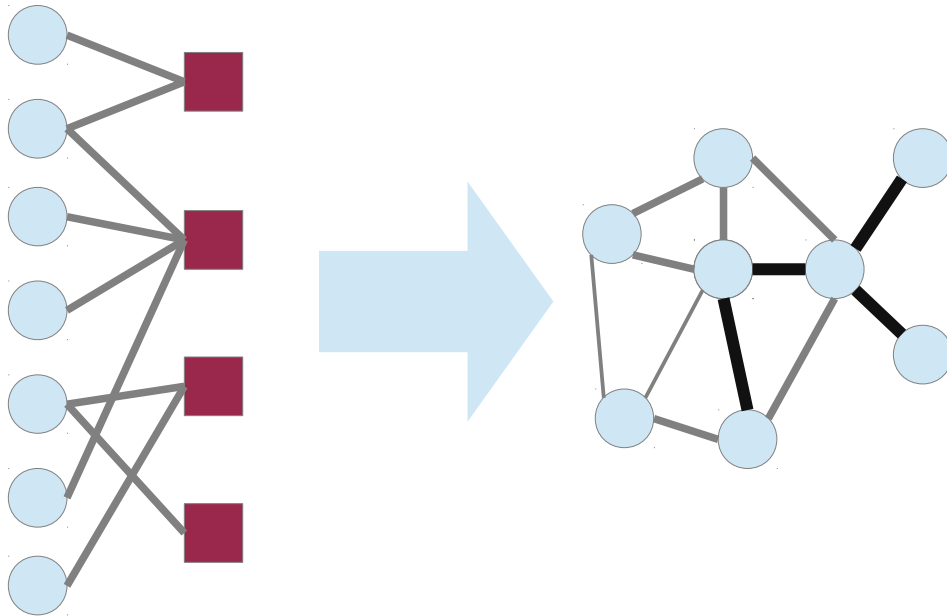
$$\sigma_u^2 = f_u (1 - f_u)$$

Methodology



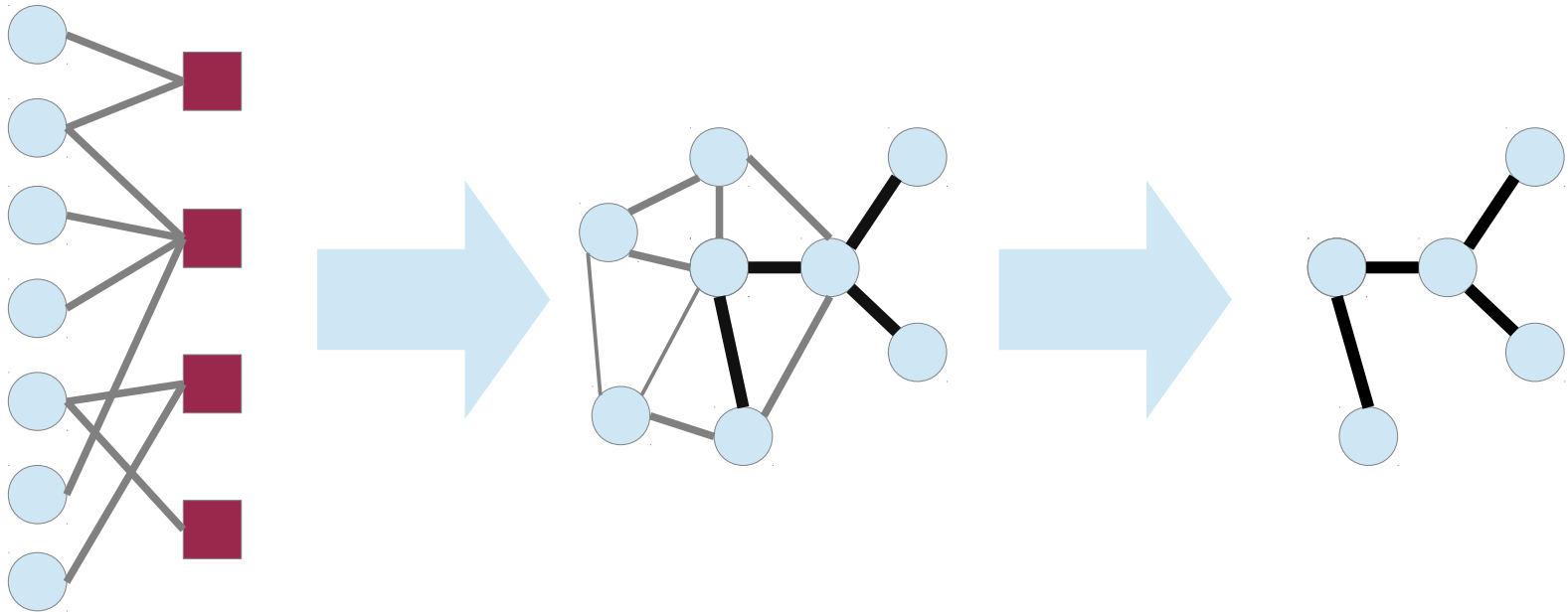
COLLECT

Methodology



PROJECT

Methodology



SELECT