

Structural measures for multiplex networks



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- General formalism for multiplex networks

- Measures to characterise the multiplexity of a system
 - 1 Basic node and link properties
 - 2 Local properties (clustering)
 - 3 Global properties (transitivity, reachability, centrality)

- Validation of all measures on a genuine multi-layer dataset of Indonesian terrorists

General formalism for multiplex networks



A multiplex is a system whose basic units are connected through a variety of different relationships. Links of different kind are embedded in different layers.

- Node index $i = 1, \dots, N$
- Layer index $\alpha = 1, \dots, M$

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For each layer α :

- adjacency matrix $A^{[\alpha]} = \{a_{ij}^{[\alpha]}\}$
- node degree $k_i^{[\alpha]} = \sum_j a_{ij}^{[\alpha]}$
- $\sum_i k_i^{[\alpha]} = 2K^{[\alpha]}$ $K^{[\alpha]}$ is the size of layer α

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For the multiplex:

- vector of adjacency matrices $\mathbf{A} = \{A^{[1]}, \dots, A^{[M]}\}$.
- vector of degrees $\mathbf{k}_i = (k_i^{[1]}, \dots, k_i^{[M]})$.

Vectorial variables are necessary to store all the richness of multiplexes.

General formalism: aggregated matrices

Aggregated topological network:

- adjacency matrix $\mathcal{A} = \{a_{ij}\}$:

$$a_{ij} = \begin{cases} 1 & \text{if } \exists \alpha : a_{ij}^{[\alpha]} = 1 \\ 0 & \text{otherwise} \end{cases} \quad (1)$$

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Aggregated overlapping network:

- adjacency matrix $\mathcal{O} = \{o_{ij}\}$:

$$o_{ij} = \sum_{\alpha} a_{ij}^{[\alpha]} \quad \text{edge overlap} \quad (2)$$

- node degree $o_i = \sum_j o_{ij} = \sum_{\alpha} k_i^{[\alpha]}, \quad o_i \geq k_i$
- $\sum_i o_i = 2O$

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Scalar variables describing system's multiplexity cannot disregard the layer index $[\alpha]$.

Generalisation to the case of weighted layers: $a_{ij}^{[\alpha]} \rightarrow w_{ij}^{[\alpha]} \quad k_i^{[\alpha]} \rightarrow s_i^{[\alpha]} \quad o_{ij} \rightarrow o_{ij}^w$

The multi-layer network of Indonesian terrorists

- 78 nodes

- 911 edges representing 4 social relationships:

- 1 Trust (weighted edges)

- 2 Operations (weighted edges)

- 3 Communications

- 4 Business

(only few information)

The multi-layer network of Indonesian terrorists

LAYER	CODE	N_{act}	K	S	O	O^w
MULTIPLEX	M	78	623	/	911	1014
Trust	T	70	259	293	/	/
Operations	O	68	437	506	/	/
Communications	C	74	200	200	/	/
Business	B	13	15	15	/	/

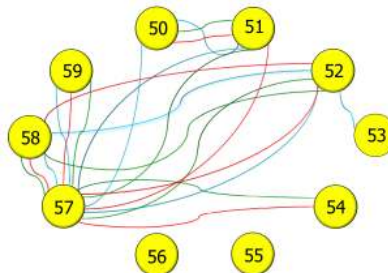
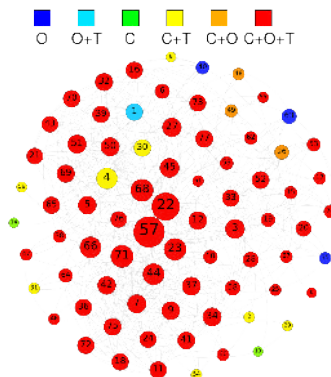


Figure : Coloured-edge representation of a subset of 10 nodes for the multiplex network of Indonesian terrorist: green edges represent trust, red edges communications and blue edges common operations.

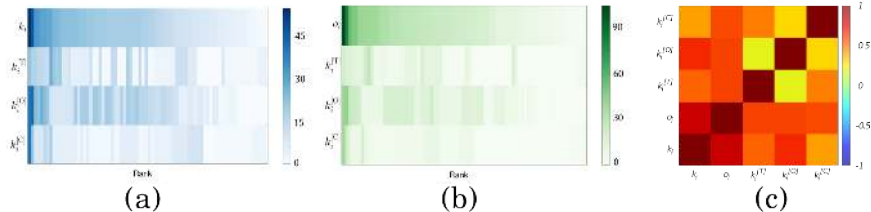
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Basic node properties

A layer-by-layer exploration of node properties: the case of the degree distribution.



Different layers show different patterns.

Basic node properties: cartography of a multiplex

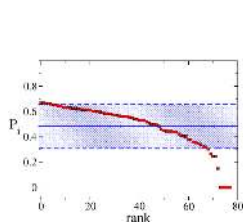
Participation coefficient:
$$P_i = 1 - \sum_{\alpha=1}^M \left(\frac{k_i^{[\alpha]}}{o_i} \right)^2$$

- 1 Focused nodes $0 \leq P_i \leq 0.3$
- 2 Mixed-pattern nodes $0.3 < P_i \leq 0.6$
- 3 Truly multiplex nodes $P_i > 0.6$

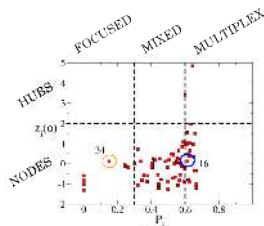
Z-score of the overlapping degree:
$$z_i(o) = \frac{o_i - \langle o \rangle}{\sigma_o}$$

- 1 Simple nodes $-2 \leq z_i(o) \leq 2$
- 2 Hubs $z_i(o) > 2$

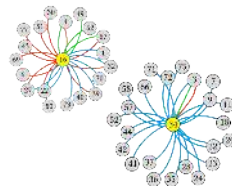
Basic node properties: cartography of a multiplex



(a)



(b)



(c)

Edge overlap and social reinforcement

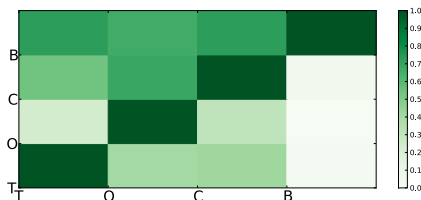
o_{ij}	Percentage of edges (%)
1	46
2	27
3	23
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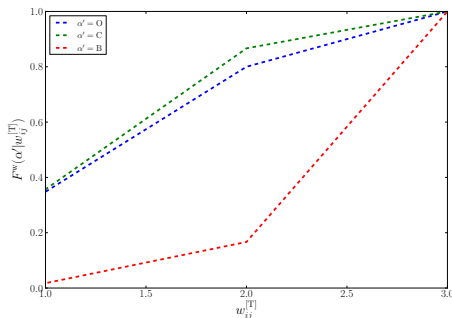
Probability conditional overlap:

$$F(a_{ij}^{[\alpha']} | a_{ij}^{[\alpha]}) = \frac{\sum_{ij} a_{ij}^{[\alpha']} a_{ij}^{[\alpha]}}{\sum_{ij} a_{ij}^{[\alpha]}} \quad (3)$$



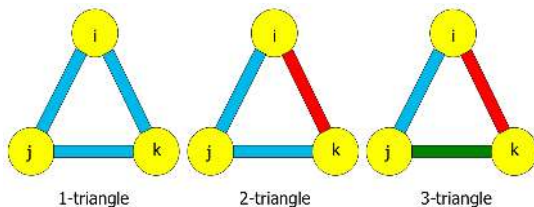
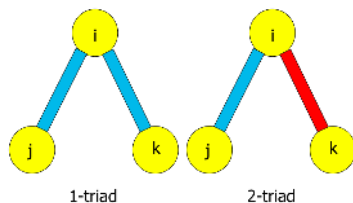
Edge overlap and social reinforcement

$$F(a_{ij}^{[\alpha']} | a_{ij}^{[\alpha]}) \rightarrow F^w(a_{ij}^{[\alpha']} | w_{ij}^{[\alpha]})$$



The existence of strong connections in the Trust layer, which represents the strongest relationships between two people, actually fosters the creation of links in other layers.

Triads and triangles

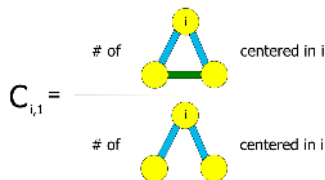


Clustering

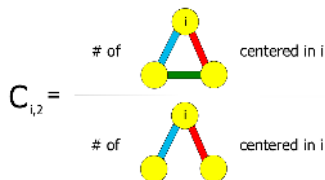
$$C_{i,1} = \frac{\begin{array}{c} \text{\# of} \\ \text{centered in } i \\ \text{\# of} \\ \text{centered in } i \end{array} \frac{\begin{array}{c} \text{triangle} \\ \text{centered in } i \end{array}}{\begin{array}{c} \text{star} \\ \text{centered in } i \end{array}}}{\begin{array}{c} \text{\# of} \\ \text{centered in } i \end{array}}$$

$$C_{i,1} = \frac{\sum_{\alpha} \sum_{\alpha' \neq \alpha} \sum_{j \neq i, m \neq i} (a_{ij}^{[\alpha]} a_{jm}^{[\alpha']} a_{mi}^{[\alpha]})}{\sum_{\alpha} \sum_{j \neq i, m \neq i} (a_{ij}^{[\alpha]} a_{mi}^{[\alpha]})} \quad (4)$$

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$$C_{i,2} = \frac{\sum_{\alpha} \sum_{\alpha' \neq \alpha} \sum_{\alpha'' \neq \alpha, \alpha'} \sum_{j \neq i, m \neq i} (a_{ij}^{[\alpha]} a_{jm}^{[\alpha'']} a_{mi}^{[\alpha']})}{\sum_{\alpha} \sum_{\alpha' \neq \alpha} \sum_{j \neq i, m \neq i} (a_{ij}^{[\alpha]} a_{mi}^{[\alpha']})} \quad (5)$$

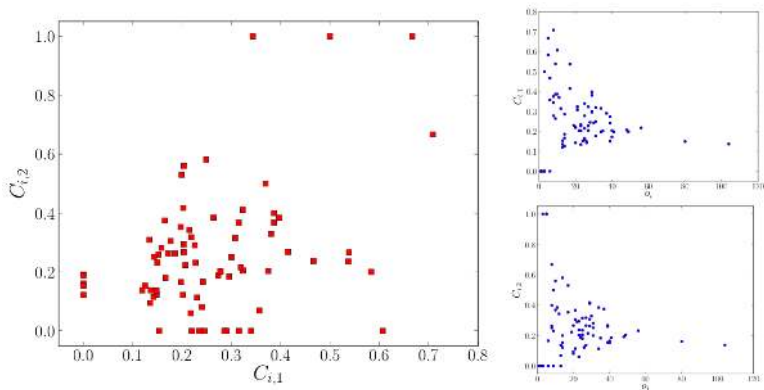
Transitivity

$$T_1 = \frac{\text{\# of } \begin{array}{c} \text{triangle} \end{array} \text{ in the multiplex}}{\text{\# of } \begin{array}{c} \text{V} \end{array} \text{ in the multiplex}}$$

$$T_2 = \frac{\text{\# of } \begin{array}{c} \text{triangle} \end{array} \text{ in the multiplex}}{\text{\# of } \begin{array}{c} \text{V} \end{array} \text{ in the multiplex}}$$

Clustering

$C_{i,1}$ and $C_{i,2}$ show different patterns of multi-clustering and are not correlated with o_i .



Transitivity and clustering

We call configuration model (CM) the set of multiplexes obtained from the original system by randomising edges and keeping fixed the sequence of degree vectors $\mathbf{k}_1, \mathbf{k}_2, \dots, \mathbf{k}_N$, i.e. keeping fixed the degree sequence at each layer α .

Variable	Real data	Randomised data
C_1	0.26	0.17
C_2	0.26	0.18
T_1	0.21	0.15
T_2	0.21	0.16

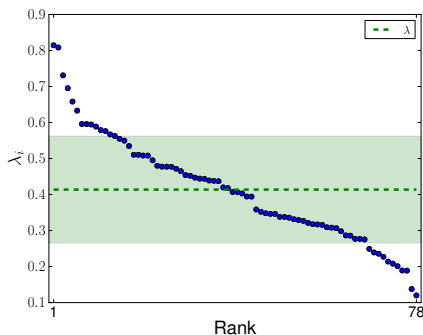
Measures of multi-clustering for real data are systematically higher than the ones obtained for randomised data, where edge correlations are washed out by randomisation

Navigability: shortest paths and interdependence

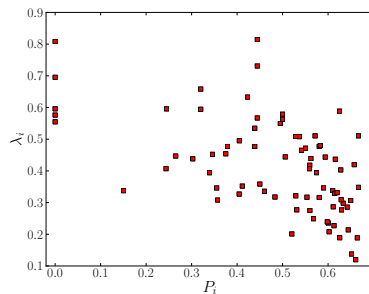
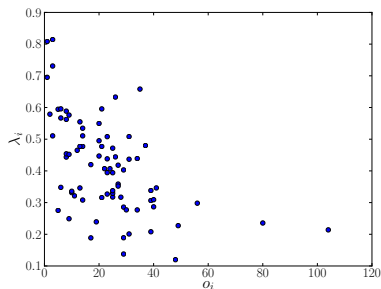
The interdependence λ_i of node i is defined as:

$$\lambda_i = \sum_{j \neq i} \frac{\psi_{ij}}{\sigma_{ij}} \quad (6)$$

where σ_{ij} is the total number of shortest paths between i and j and ψ_{ij} is the number of interdependent shortest paths between node i and node j .



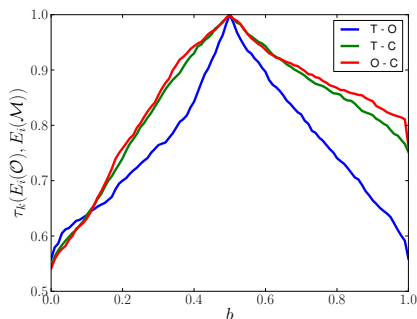
Reachability: shortest paths and interdependence



Centrality

For a duplex we can construct the following adjacency matrix:

$$\mathcal{M}(b) = bA^{[1]} + (1-b)A^{[2]}, \quad \mathcal{M}(b=0.5) = \mathcal{O} \quad (7)$$



The symmetry/asymmetry of the curves tell us about the interplay between the layers in determining the centrality of the multi-layer system.

Both layers T and O dominate C in determining the centrality of the multiplex.

Summary



- We suggested a comprehensive formalism to deal with systems composed of several layers

- We also proposed a number of metrics to characterize multiplex systems with respect to:
 - 1 Node degree
 - 2 Node participation to different layers
 - 3 Edge overlap
 - 4 Clustering
 - 5 Transitivity
 - 6 Reachability
 - 7 Eigenvector centrality

- 8 You can find more in the paper

"Structural measures for multiplex networks", *Phys. Rev. E* **89** 032804 (2014).
F. Battiston, V. Nicosia and V. Latora,
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