

core-periphery structure

introduction to *network analysis* (*ina*)

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core-periphery *block model*



community block model [GN02]



disassortative (bipartite) block model [NL07]



core-periphery block model [Sei83]

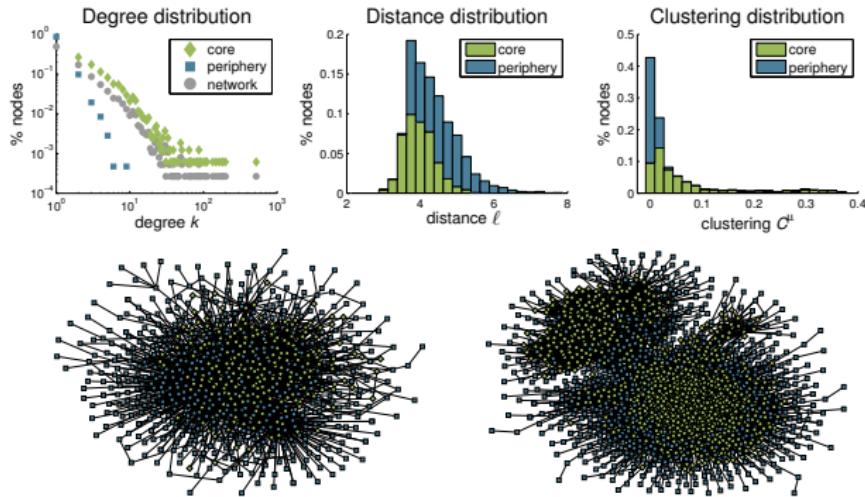


random block model [ER59]

* origin of core-periphery structure in international relations

core-periphery *structure*

- core/periphery nodes have higher/lower degrees k
- core/periphery nodes are on shorter/longer distances ℓ
- core/periphery nodes have higher/lower clustering C

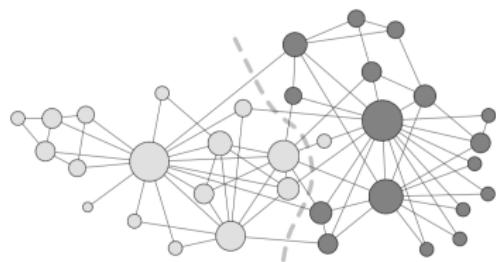


core-periphery *stochastic*

- $G(\{C_1, C_2\}, \{p_{11}, p_{12}, p_{22}\})$ stochastic block model [HLL83]
 - n_i is size of cluster C_i & p_{ij} is link density between C_i and C_j
- density-based core-periphery structure when $p_{11} \gg p_{12} \gg p_{22}$
- lookalike core-periph. when $n_1 p_{11} \gg 1$, $n_1 p_{12} \ll 1$, $n_2 p_{22} \approx 1$



non-corrected block model $p_{11} > p_{12} > p_{22}$



degree-corrected block model $p_{11} \approx p_{22} > p_{12}$

core-periphery *discrete/continuos*

- discrete core-periphery division $\delta \in \{0, 1\}$ [BE00]

– $\delta_i = 1$ for core nodes i & $\delta_i = 0$ for peripheral nodes i

$$\rho_{\{0,1\}} = \sum_{ij} A_{ij} \Delta_{ij} \quad \Delta_{ij} = \begin{cases} 1 & \text{if } \delta_i = \delta_j = 1 \\ 0 & \text{if } \delta_i = \delta_j = 0 \\ \in [0, 1] & \text{if } \delta_i - \delta_j \neq 0 \end{cases}$$

- continuos core-periphery centrality $\delta \in [0, 1]$

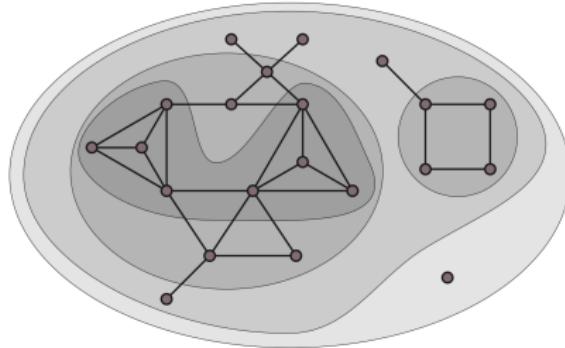
– $\delta_i \approx 1$ for core nodes i & $\delta_i \approx 0$ for peripheral nodes i

$$\rho_{[0,1]} = \sum_{ij} A_{ij} \delta_i \delta_j$$

$$\Delta^1 = \left[\begin{array}{ccc|ccc} 0 & 1 & 1 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 & 1 & 1 \\ 1 & 1 & 0 & 1 & 1 & 1 \\ \hline 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \\ 1 & 1 & 1 & 0 & 0 & 0 \end{array} \right] \quad \Delta^\alpha = \left[\begin{array}{ccc|ccc} 0 & 1 & 1 & \alpha & \alpha & \alpha \\ 1 & 0 & 1 & \alpha & \alpha & \alpha \\ 1 & 1 & 0 & \alpha & \alpha & \alpha \\ \hline \alpha & \alpha & \alpha & 0 & 0 & 0 \\ \alpha & \alpha & \alpha & 0 & 0 & 0 \\ \alpha & \alpha & \alpha & 0 & 0 & 0 \end{array} \right] \quad \delta = \begin{bmatrix} 1 \\ 0.8 \\ 0.7 \\ 0.4 \\ 0.2 \\ 0.1 \end{bmatrix}$$

core-periphery k -cores

- k -cores are *subgraphs of nodes with $\geq k$ neighbors* [Sei83]
remove nodes with degree $< k$ until no such node remains [BZ11]
- k -shells are *nodes of k -cores that are not in $k+1$ -cores*
- k -cores are *nested* while k -shells form *decomposition*



1-cores are connected components w/o isolates & k -cores can be disconnected

core-periphery k^* -core

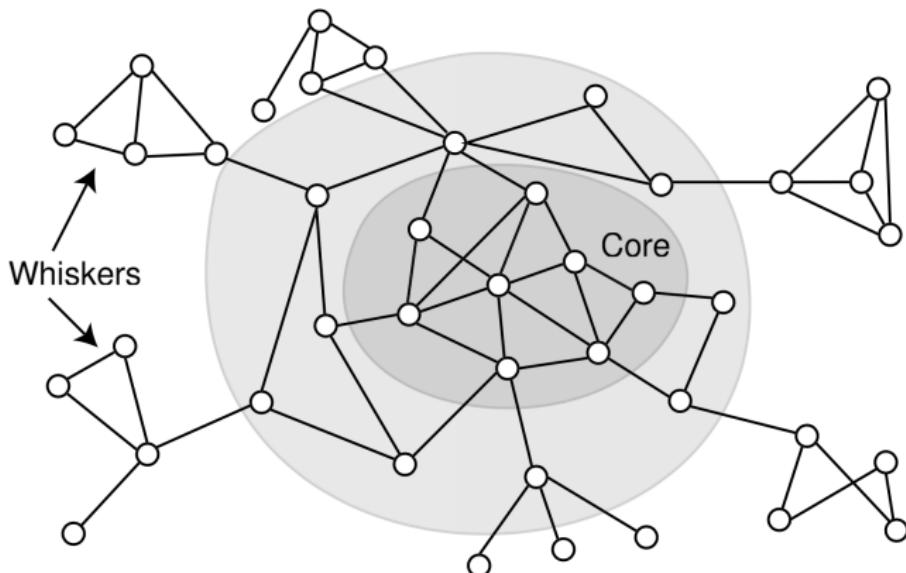
- Holme's k^* -core maximizes closeness centrality ℓ^{-1} [Hol05]
 - d_{ij} is distance between i and j & ℓ_i is farness centrality of i
 - ℓ_C^{-1} is closeness centrality of cluster C & n_c is size of C
- Holme's core-periphery coefficient c_{cp} for k^* -core
 - N is set of nodes & N_k are nodes in k -core
 - $\langle \dots \rangle_{G'}$ is expectation in random graph G'

$$c_{cp} = \ell_{N_{k^*}}^{-1} / \ell_N^{-1} - \left\langle \ell_{N'_{k^*}}^{-1} / \ell_{N'}^{-1} \right\rangle_{G'}$$

core-periphery coefficient

Network		<i>N</i>	<i>M</i>	<i>c_{cp}</i>
Geographical networks	Interstate highways	935	1315	0.231(1)
	Pipelines	2999	3079	0.180(2)
	Streets, Stockholm	3325	5100	0.255(1)
	Streets, Göteborg	1258	1516	0.040(3)
	Airport	449	2795	0.0523(3)
	Internet	1968(66)	4051(121)	0.045(2)
One-mode projections of affiliation networks	arXiv	48561	287570	-0.08(3)
	Board of directors	6193	43074	-0.037(2)
	Ajou University students	7285(128)	75898(6566)	-0.08(1)
Acquaintance networks	High School friendship	571(43)	1078(85)	0.006(7)
	Prisoners	58	83	-0.043(2)
	Social scientists	34	265(35)	-0.002(4)
Electronic communication	e-mail, Ebel <i>et al.</i>	39592	57703	-0.229(4)
	e-mail, Eckmann <i>et al.</i>	3186	31856	-0.091(2)
	Internet community, nioki.com	49801	239265	-0.014(2)
	Internet community, pussokram.com	28295	115335	-0.183(5)
Reference networks	WWW, nd.edu	325729	1090108	-0.027(3)
	HEP citations	27400	352021	-0.10(1)
Software dependencies	GNU / Linux	504	793	-0.155(1)
Food webs	Little Rock Lake	92	960	0.005(6)
	Ythan Estuary	134	593	-0.020(1)
Neural network	<i>C. elegans</i>	280	1973	0.040(6)
Biochemical networks	<i>Drosophila</i> protein	2915	4121	-0.035(2)
	<i>S. cerevisiae</i> protein	3898	7283	-0.249(1)
	<i>S. cerevisiae</i> genetic	1503	5043	-0.0646(7)
	Metabolic networks	427(27)	1257(88)	-0.002(6)
	Whole cellular networks	623(32)	1752(103)	-0.004(6)

core-periphery *nestedness*



nested cores & whiskers communities [LLDM09, YL13]

core-periphery *references*

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core-periphery *references*



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