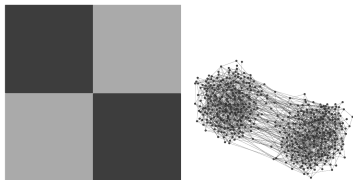


## *core-periphery* structure

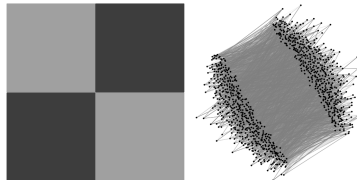
introduction to *network analysis in Python* (*NetPy*)

Lovro Šubelj  
University of Ljubljana  
19th Sep 2019

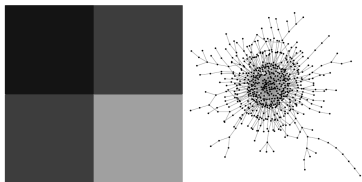
# core-periphery *block model*



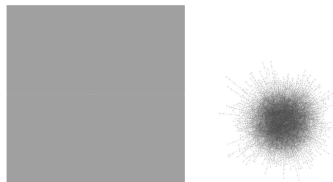
*community block model* [GN02]



*disassortative block model* [NL07]



*core-periphery block model* [Sei83]



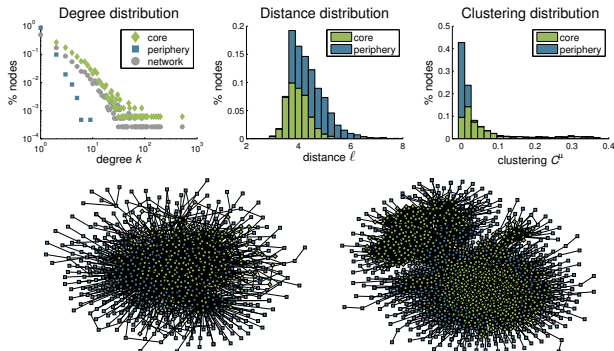
*random graph* [ER59]

---

\* origin of core-periphery structure in international relations

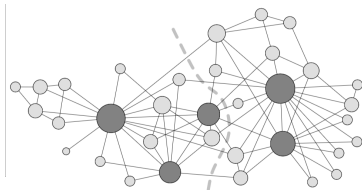
# core-periphery *structure*

- *core/periphery nodes* have *higher/lower degrees*  $k$
- *core/periphery nodes* are on *shorter/longer distances*  $\ell$
- *core/periphery nodes* have *higher/lower clustering*  $C$

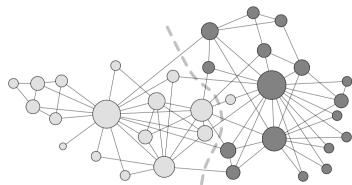


## core-periphery *stochastic*

- $G(\{C_1, C_2\}, \{p_{11}, p_{12}, p_{22}\})$  *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 for  $p_{11} \gg p_{12} \gg p_{22}$
- *lookalike core-periphery* for  $n_1 p_{11} \gg 1$ ,  $n_1 p_{12} \ll 1$ ,  $n_2 p_{22} \approx 1$



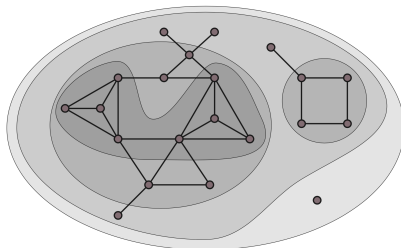
*non-corrected block model*  $p_{11} > p_{12} > p_{22}$



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

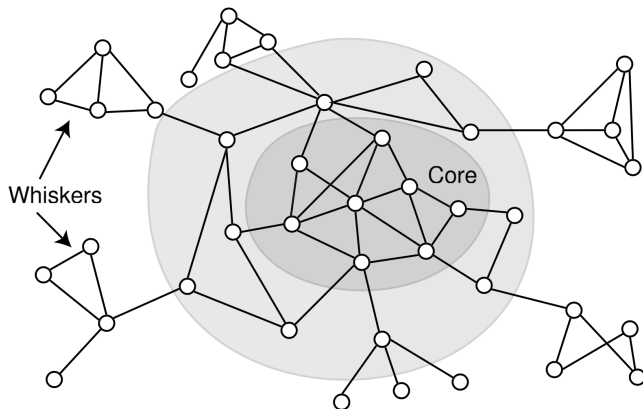
## core-periphery *k*-cores

- *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



0-cores are connected components & *k*-cores can be disconnected

## core-periphery *nestedness*



*nested cores & whiskers communities* [LLDM09, YL13]

# core-periphery *references*



V. Batagelj and M. Zaveršnik.

An  $O(m)$  algorithm for cores decomposition of networks.  
*Adv. Data Anal. Classif.*, 5(2):129–145, 2011.



P. Erdős and A. Rényi.

On random graphs I.  
*Publ. Math. Debrecen*, 6:290–297, 1959.



M. Girvan and M. E. J Newman.

Community structure in social and biological networks.  
*P. Natl. Acad. Sci. USA*, 99(12):7821–7826, 2002.



Paul W. Holland, Kathryn Blackmond Laskey, and Samuel Leinhardt.

Stochastic blockmodels: First steps.  
*Soc. Networks*, 5(2):109–137, 1983.



Jure Leskovec, Kevin J Lang, Anirban Dasgupta, and Michael W Mahoney.

Community structure in large networks: Natural cluster sizes and the absence of large well-defined clusters.  
*Internet Math.*, 6(1):29–123, 2009.



Tilen Marc and Lovro Šubelj.

Convexity in complex networks.  
*Netw. Sci.*, 6(2):176–203, 2018.



M. E. J Newman and E. A Leicht.

Mixture models and exploratory analysis in networks.  
*P. Natl. Acad. Sci. USA*, 104(23):9564–9569, 2007.



Stephen B. Seidman.

Network structure and minimum degree.  
*Soc. Networks*, 5(3):269–287, 1983.

## core-periphery *references*



J. Yang and Jure Leskovec.

Overlapping community detection at scale: A nonnegative matrix factorization approach.

In *Proceedings of the ACM International Conference on Web Search and Data Mining*, pages 587–596, Rome, Italy, 2013.