



Social Networks



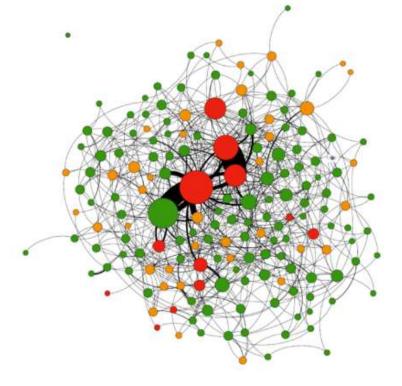
 A social network can be constructed from relational data and can be defined as a set of social entities, such as people, groups, and organizations, with some relationships or interactions between them. These networks are usually modelled by graphs, where vertices represent the social entities and edges represent the relationships established between them



What is SNA?



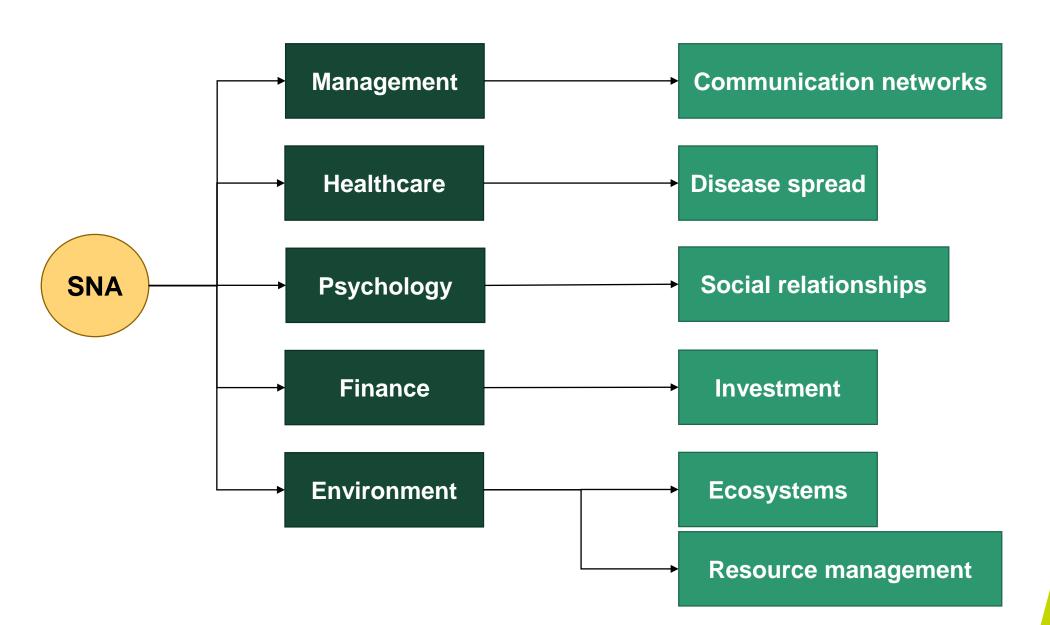
• Social network analysis studies structures of relationships linking individuals and interdependencies in behaviour or attitudes related to configurations of social relations





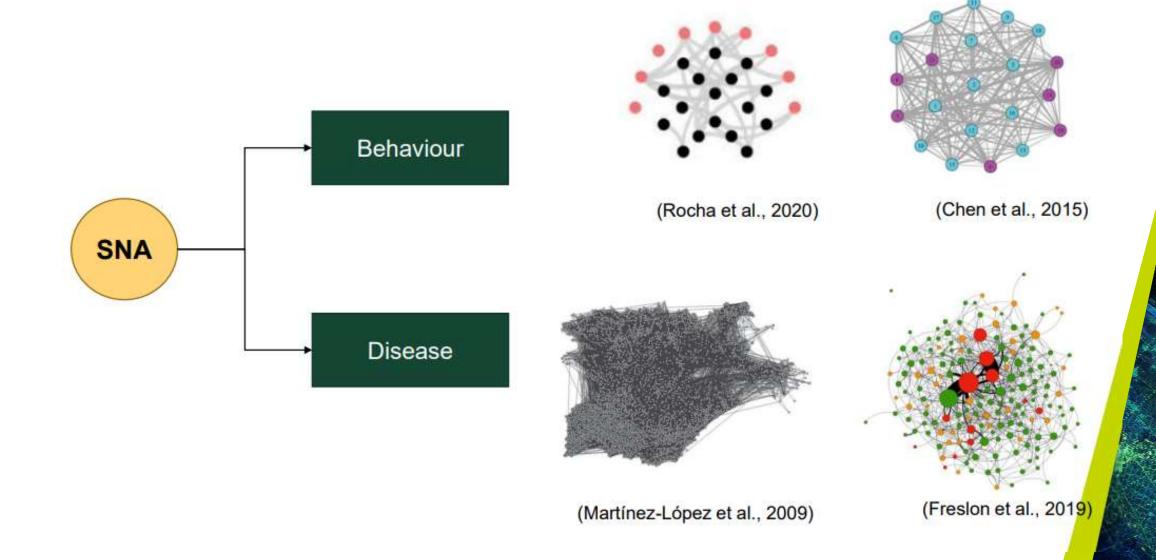
SNA applications





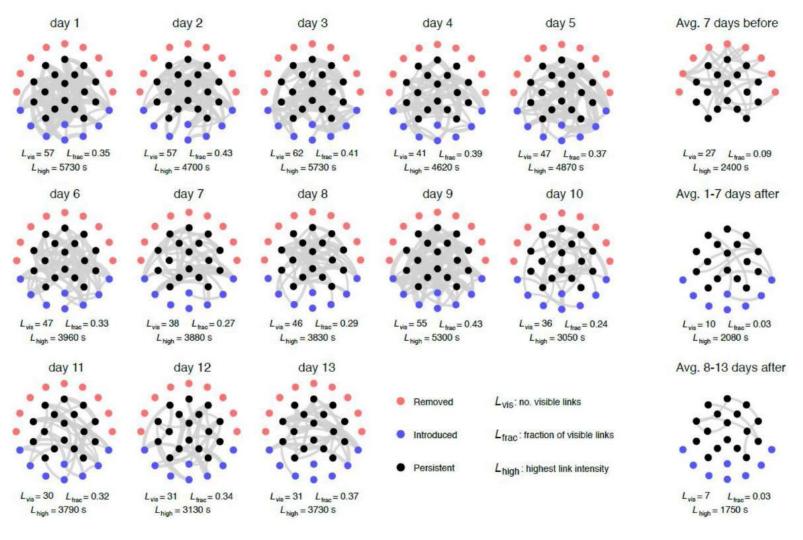










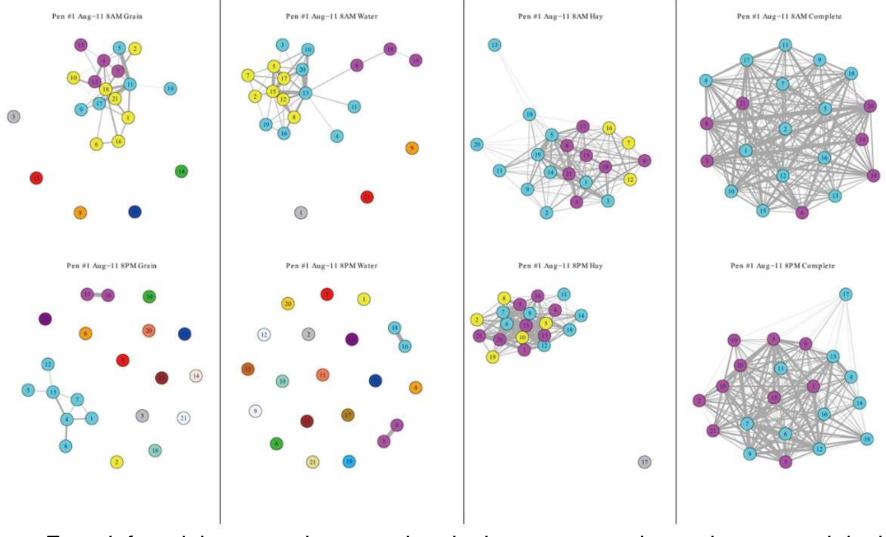


Suggested that each cow has its own sociality independent of the group and tend to establish relations with specific partners when the population is fixed

(Rocha et al., 2020)



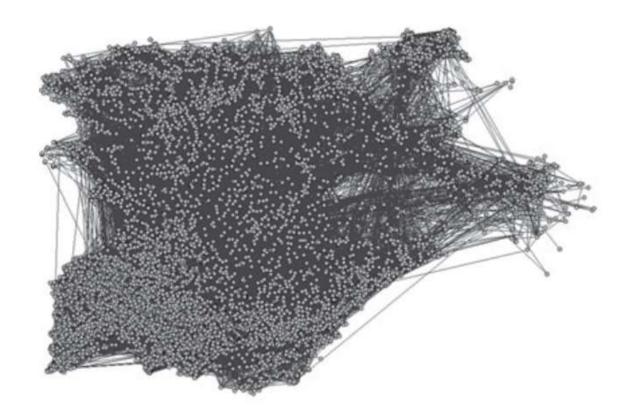




From left to right: networks around grain, hay, water, and complete network in the pen (Chen et al., 2015)



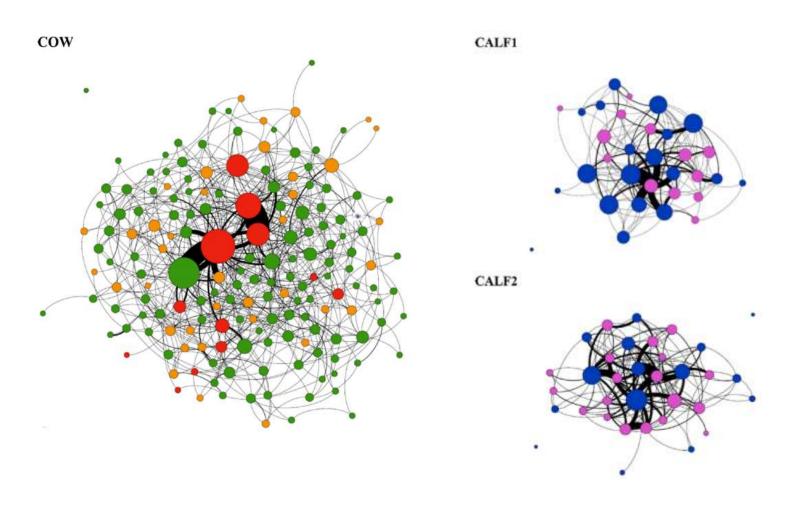




Network of cattle movements in the Spanish region of CyL in 2005. **SNA** offers important benefits for the assessment of epidemiological conditions, factors and forces associated with risk for animal disease spread







Focused on three contact behaviours that may lead to transmission of pathogenic Leptospira spp.: sniffing, licking and rubbing the face on the genital area of another animal

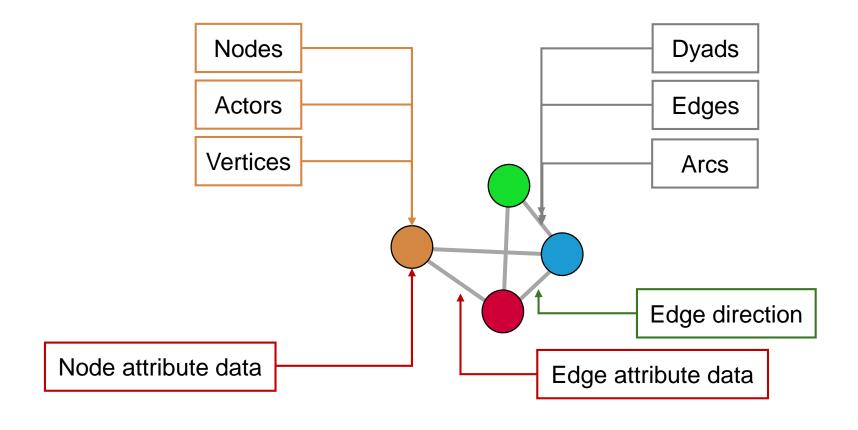


What is a network?



Part of the networks







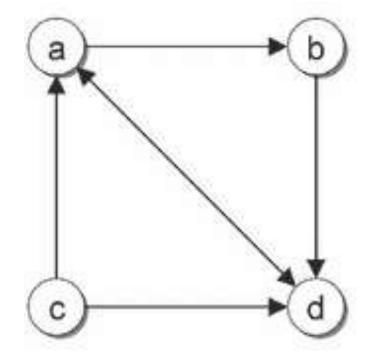
Ways to represent networks



Adjacency matrix

	а	b	С	d	
а	0	1	0	1	
b	0	0	0	1	
С	1	0	0	1	
d	1	0	0	0	

Graph



Notation

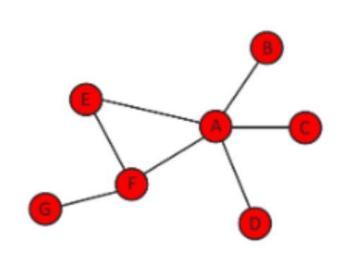
$$G = \{(a, b), (a, d), (b, d), (c, a), (c, d), (d, a)\}$$



Ways to represent networks



Α	В
Α	С
Α	D
Α	E
Α	F
Ε	F
F	G



	Α	В	С	D	E	F	G
Α	0	1	1	1	1	1	0
В	1	0	0	0	0	0	0
C	1	0	O	0	0	0	0
D	1	0	0	0	0	0	0
E	1	0	0	0	O	1	0
F	1	0	0	0	1	O	1
G	0	0	0	0	0	1	0



Software



Network analysis software





Cytoscape









sna

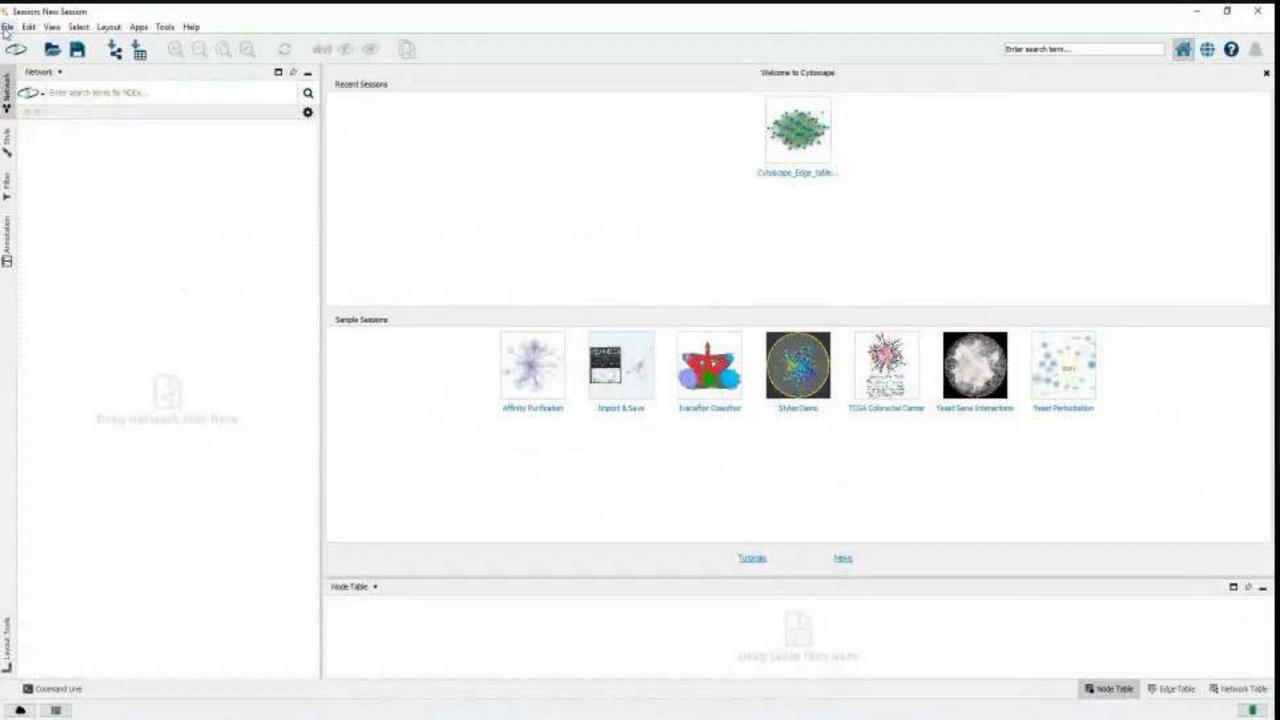
igraph

network



networkX

igraph





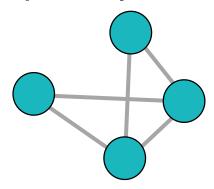


Ways to analyze the information



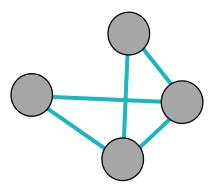
individual- level models

focuses on an individual-level outcome, network data are used to define explanatory variables



relational-level models

focuses on an **dyad-level**, analyse the **relationship** rather than a characteristic of particular individuals





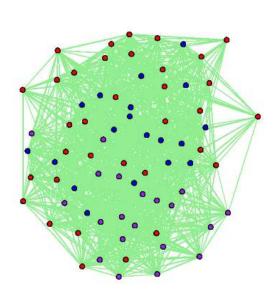
Network dimensions

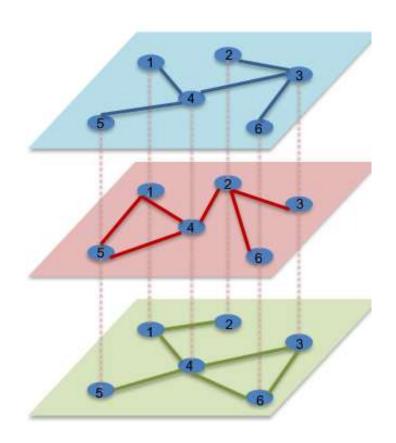


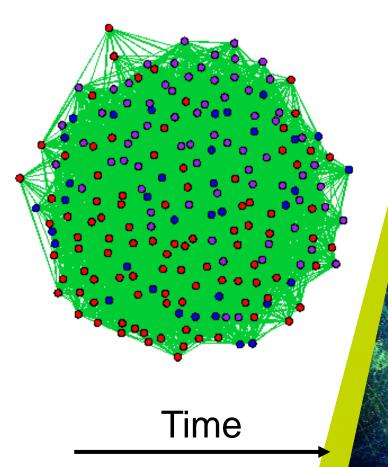
Unidimensional data

Multidimensional data

Longitudinal data

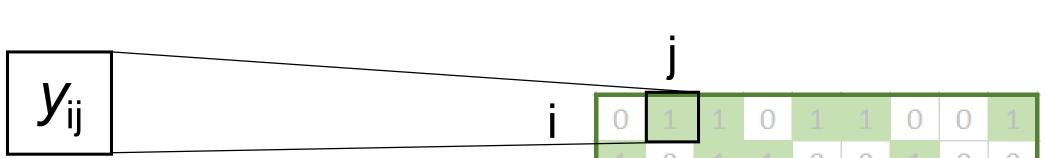




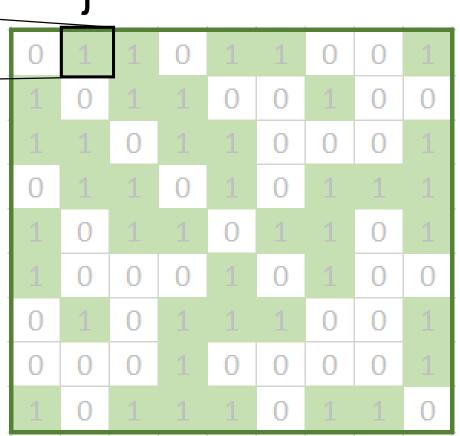








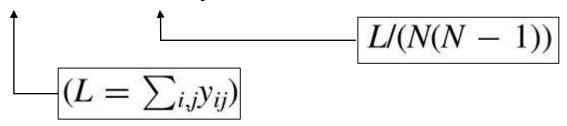
- Y is the adjacency matrix
- y_{ii}: relationship from actor i to actor j
- Binary-valued
- Self-relations are usually undefined



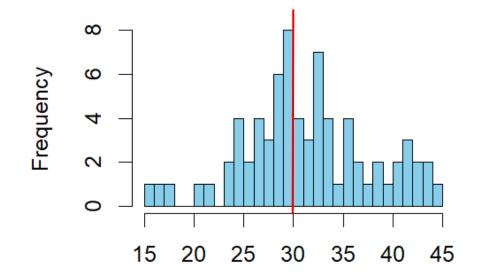


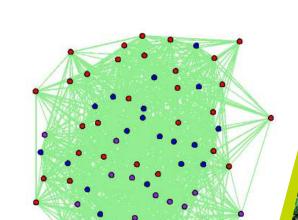


1) Size and density of the network

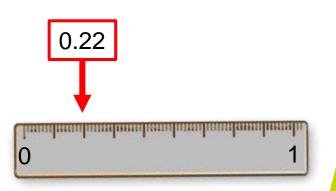


2) Degree and the degree distribution





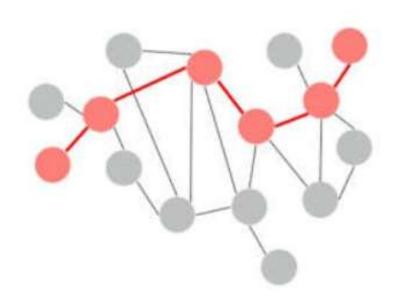
Feeding area

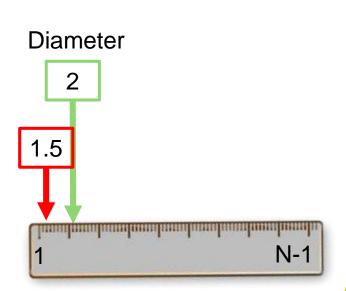






3) Geodesic distance:





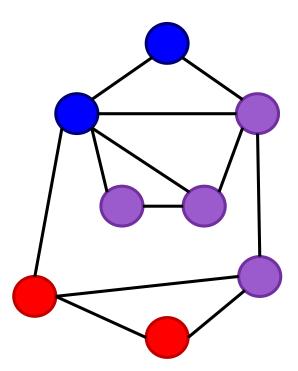




4) Centrality parameters:

Degree:

- The simplest is based on an actor's degree
- Reflects an actor's level of network activity or involvement







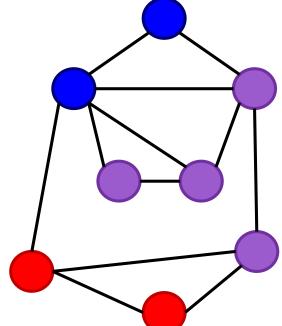
4) Centrality parameters:

Betweenness:

 Number of times a node acts as a bridge along the shortest path between two other nodes

Shortest path from s->t that cross through v

$$g(v) = \sum_{s
eq v
eq t} rac{\sigma_{st}^{\downarrow}(v)}{\sigma_{st}}$$
Shortest path from s->t





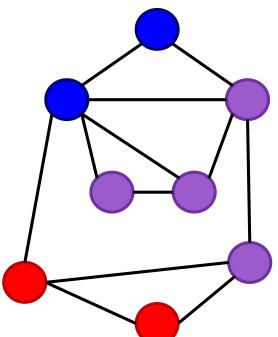


4) Centrality parameters:

Closeness:

 Sum of the length of the shortest paths between the node and all other nodes in the graph

$$C(v) = \frac{N-1}{\sum_u d(u,v)}. \begin{tabular}{ll} \longleftarrow & \text{Number of nodes in the graph} \\ \longleftarrow & \text{Distance between vertices u and v} \\ \end{tabular}$$



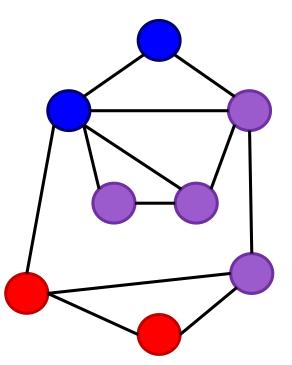




4) Centrality parameters:

Eigenvector:

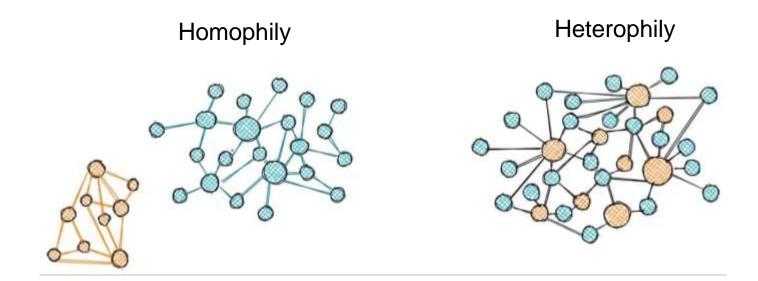
- Principal eigenvector using the adjacency matrix
- Measures a node's importance while giving consideration to the importance of its neighbors







- 5) Homophily:
- Represents the propensity of individuals to interact with others of similar characteristics



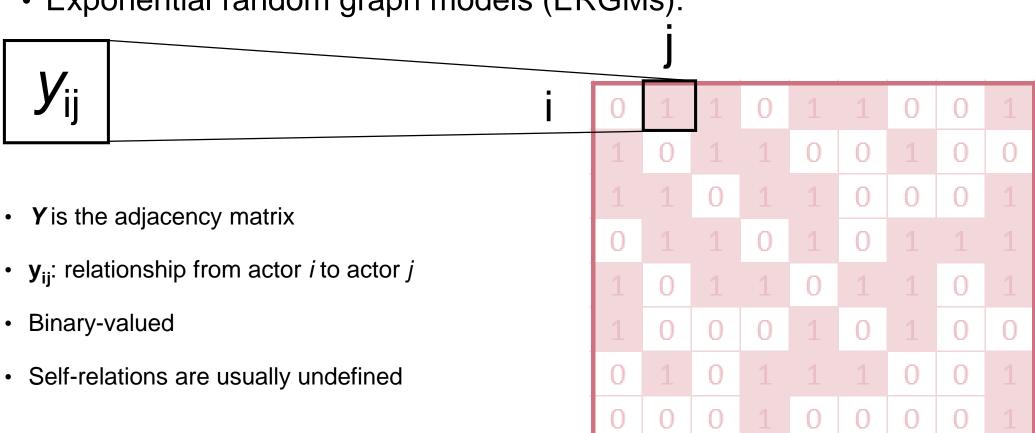


Dyad-level models





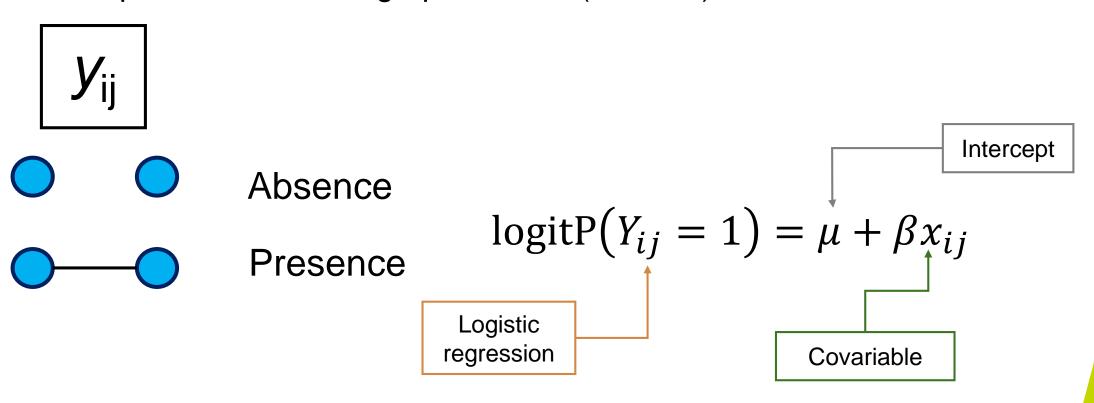
• Exponential random graph models (ERGMs):





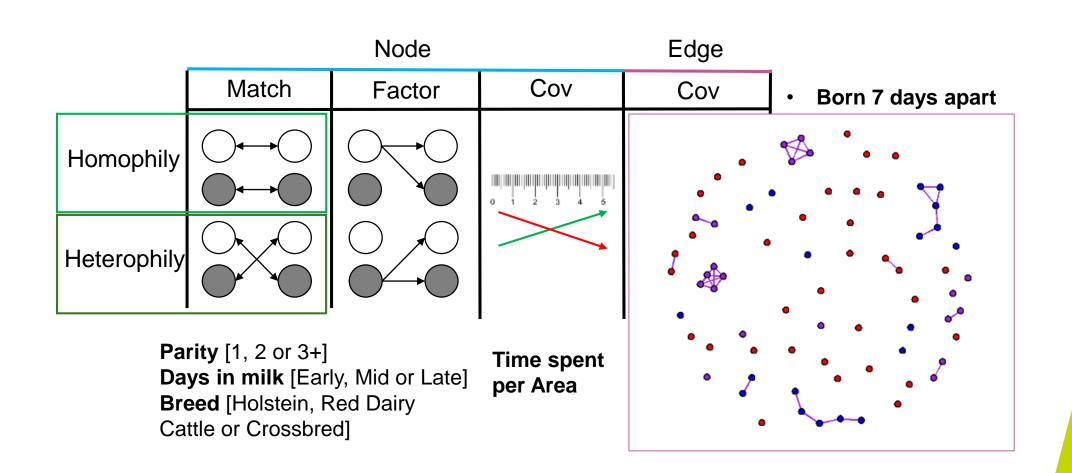


• Exponential random graph models (ERGMs):













• Exponential random graph models (ERGMs):

	Pa	rity	TimeinArea	AGENET
	Match	Factor	Cov	Cov
$ Y_{ij} = 0$	1	0 0	0.22+0.43	1
	0	0 1	0.22+0.33	0
$Y_{jk} = 1$	0	1 1	0.56+0.33	0
$Y_{im} = 1$	1	0 0	0.22+0.13	0
			•	

Parity

1

0 2

9 3+

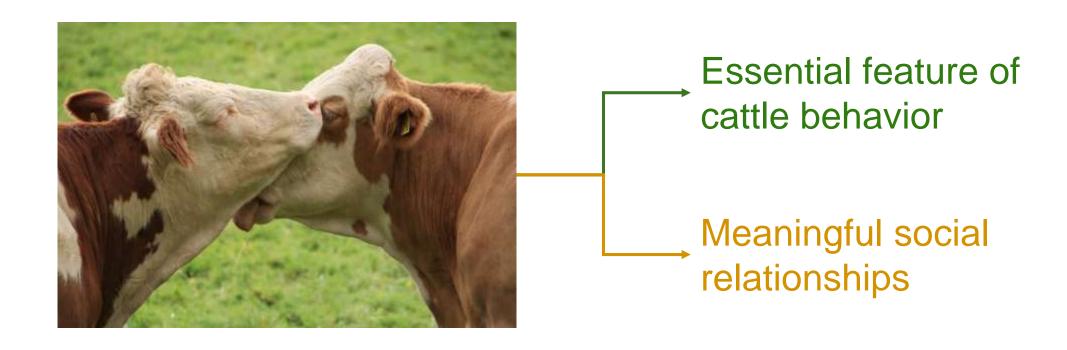


Animal information



Social interactions







Social interactions



Ultra-Wide Band technology

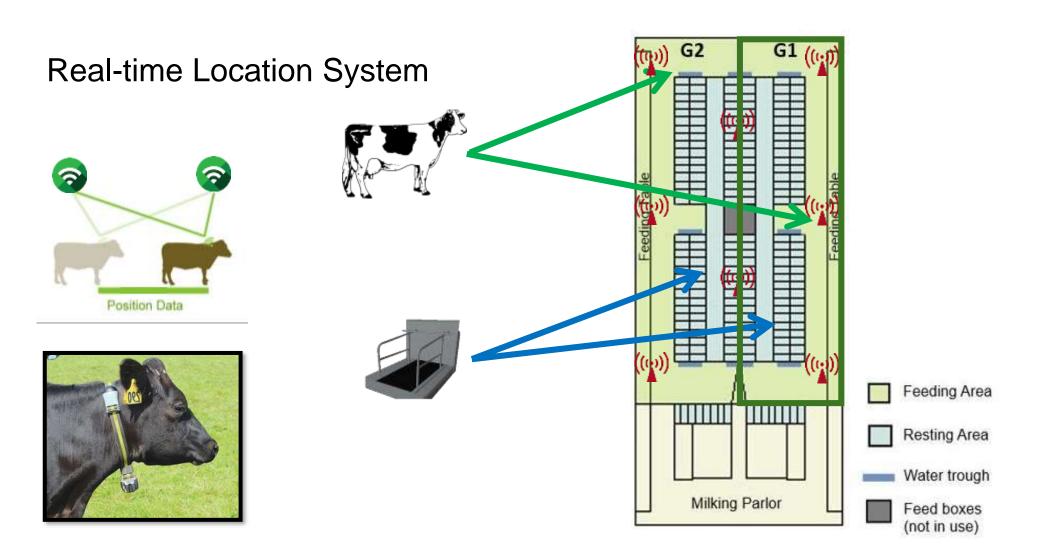


- Collecting positions of all cows every second
- Spatial interactions
- Real time information



Spatial interactions



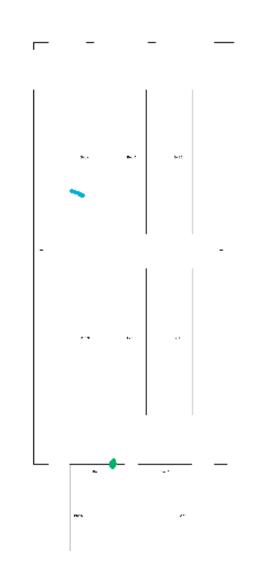


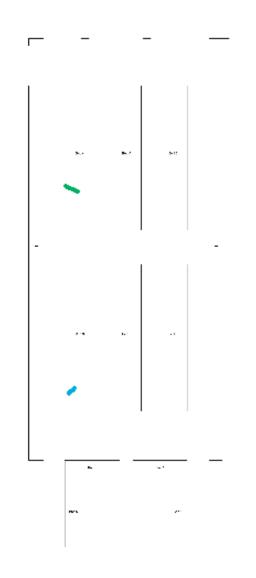
(Hansson et al., 2023)



Spatial contacts











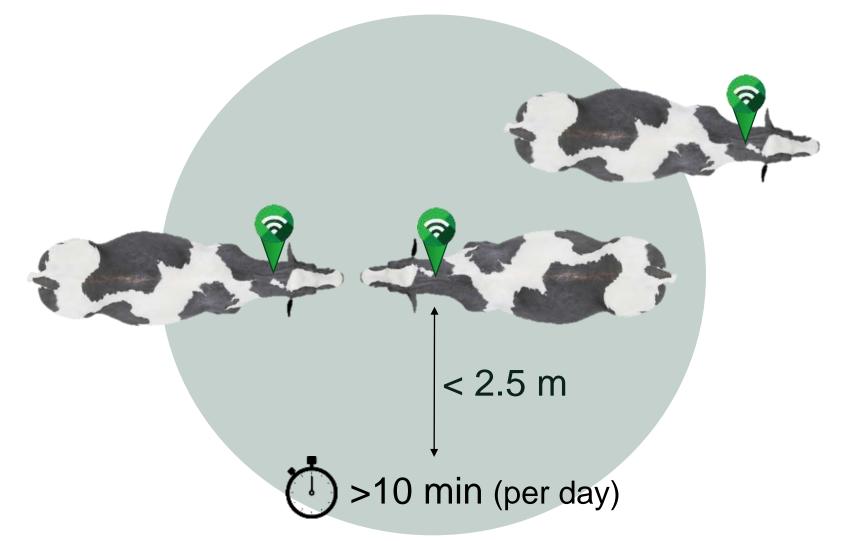




Spatial contacts



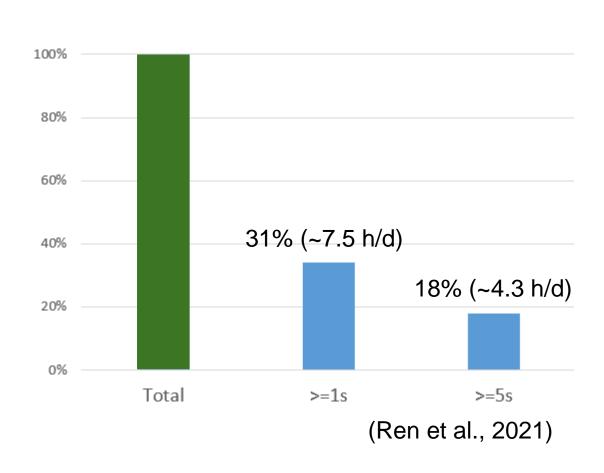
What was consider as social contact?

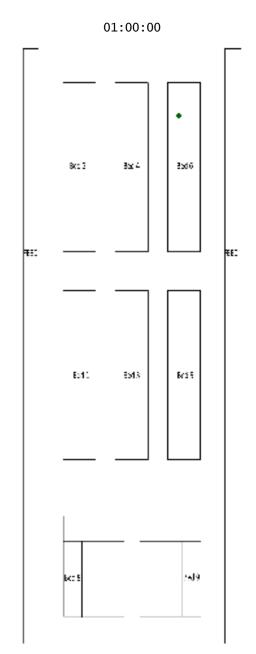




Interpolation



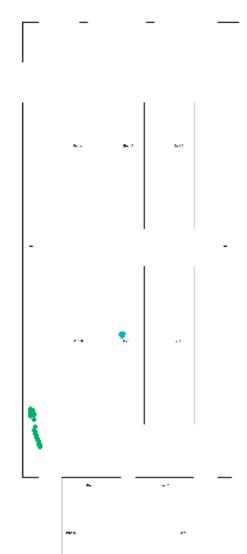


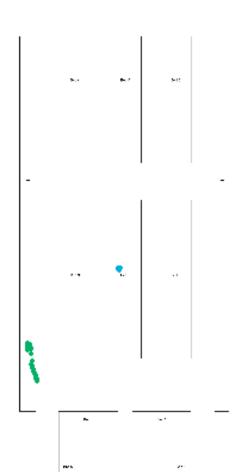


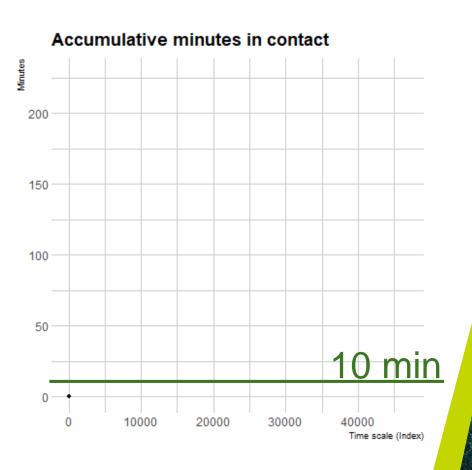


Spatial contacts











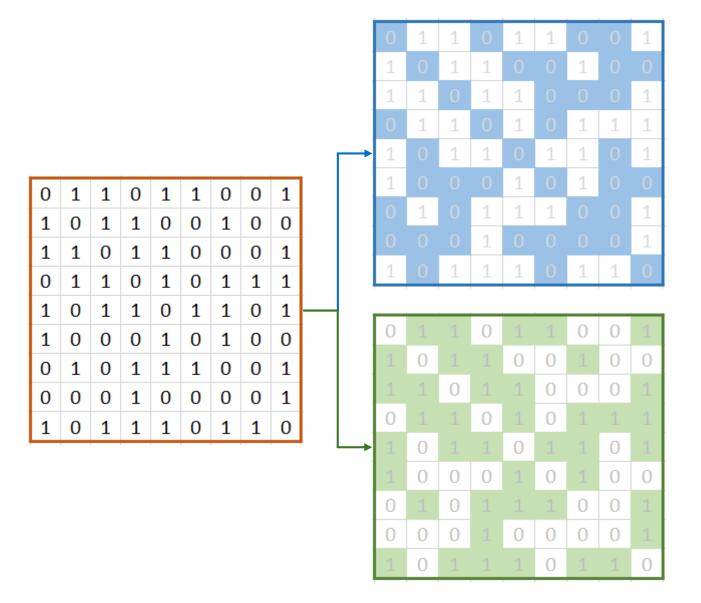






Spatial contacts





Absence

Presence









Recommended literature



- 1. O'Malley, A.J., and P. V. Marsden. 2008. The Analysis of Social Networks. Health Serv. Outcomes Res. Methodol. 8:222. doi:
- https://doi.org/10.1007/S10742-008-0041-Z
- 2. de Freslon, I., Martínez-López, B., Belkhiria, J., Strappini, A., Monti, G., 2019. Use of social network analysis to improve the understanding of social behaviour in dairy cattle and its impact on disease transmission. Appl. Anim. Behav. Sci. 213, 47–54. https://doi.org/10.1016/J.APPLANIM.2019.01.006
- 3. Tabassum, S, Pereira, FSF, Fernandes, S, Gama, J. Social network analysis: An overview. WIREs Data Mining Knowl Discov. 2018; 8:e1256. https://doi.org/10.1002/widm.1256
- 4. Martínez-López, B., Perez, A.M. and Sánchez-Vizcaíno, J.M. (2009), Social Network Analysis. Review of General Concepts and Use in Preventive Veterinary Medicine. Transboundary and Emerging Diseases, 56: 109-120. https://doi.org/10.1111/j.1865-1682.2009.01073.x
- 5. Chen, S., Ilany, A., White, B.J., Sanderson, M.W., Lanzas, C., 2015. Spatial-Temporal Dynamics of High-Resolution Animal Networks: What Can We Learn from Domestic Animals? PLoS One 10, e0129253. https://doi.org/10.1371/JOURNAL.PONE.0129253
- 6. Rocha, L.E.C., Terenius, O., Veissier, I., Meunier, B., Nielsen, P.P., 2020. Persistence of sociality in group dynamics of dairy cattle. Appl. Anim. Behav. Sci. 223, 104921. https://doi.org/10.1016/J.APPLANIM.2019.104921
- 7. Hansson, I., Silvera, A., Ren, K., Woudstra, S., Skarin, A., Fikse, W.F., Nielsen, P.P., Rönnegård, L., 2023. Cow characteristics associated with the variation in number of contacts between dairy cows. J. Dairy Sci. 106, 2685–2699. https://doi.org/10.3168/JDS.2022-21915

