







## **Between solidarity and expediency: Uncovering framing-based mechanisms** of advice network formation through an empirical agent-based model

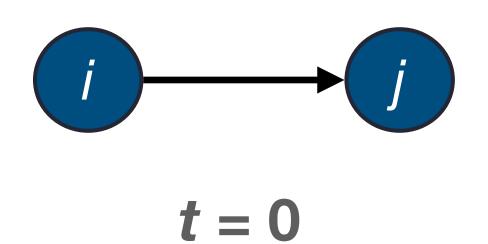
Sunbelt 2025 24 June, 2025, Sorbonne University / SciencesPo

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## **Prosocial behaviour and framing**

- Prosocial behaviour is driven by a mix of instrumental expediency and normative compliance with solidaristic obligations towards others (Simpson & Willer, 2015), which are time-varying and contextdependent (Lindenberg, 1998, 2006; Kroneberg, 2014; Esser & Kroneberg, 2015) according to actors' framing of the relationship as solidary or instrumental (Fiske, 1991)
- Ego's framing of their relationship with alter may vary over time as a macro-micro feedback of certain contextual features, such as the connectivity of the wider social network (Marwell et al., 1988; Coleman, 1988, 1991)
- Advice-seeking networks are usually found to be driven by direct reciprocation and transitive closure (e.g., Agneessens & Wittek, 2012)

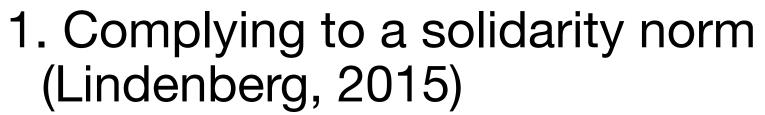


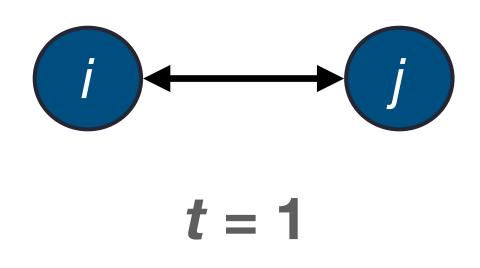


(Lindenberg, 2015)

2. Strategically investing in a longterm relationship (Coleman, 1991)





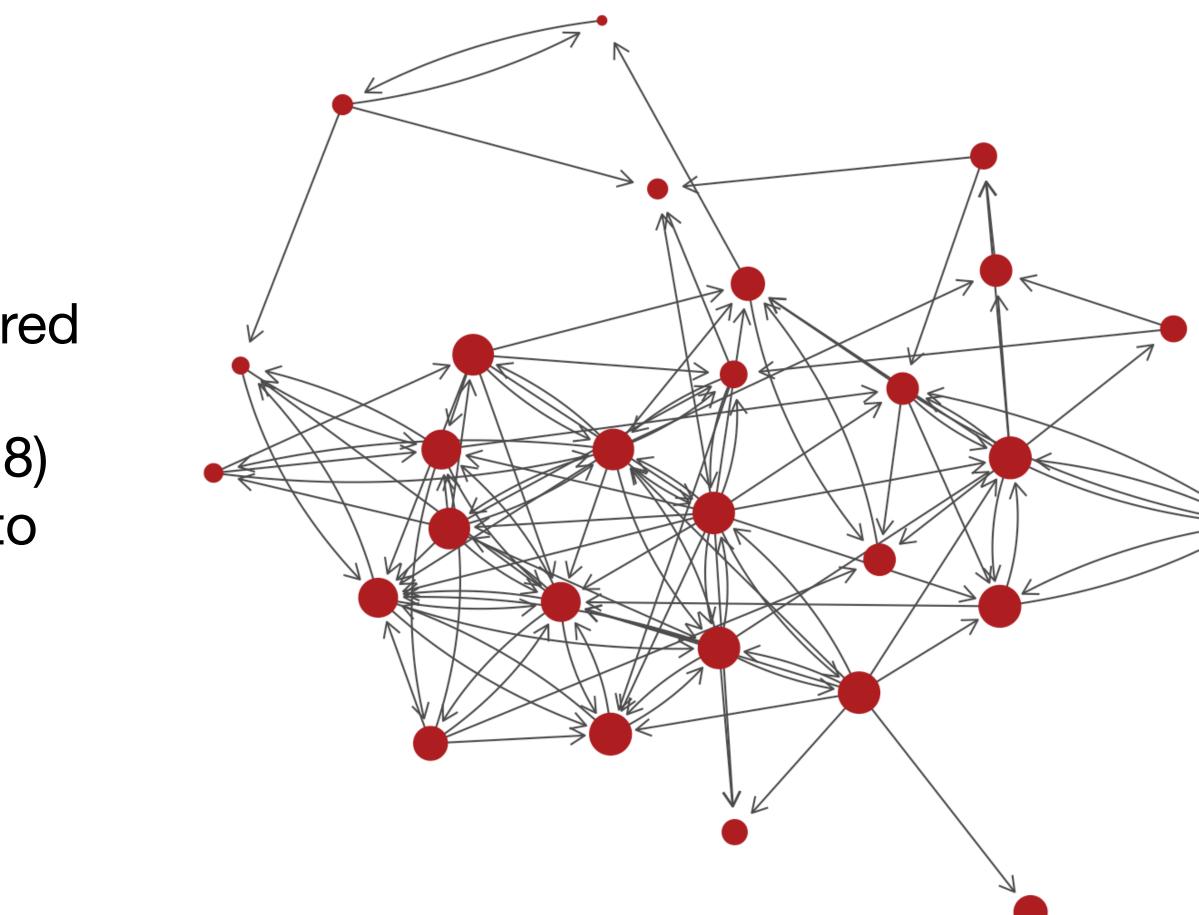




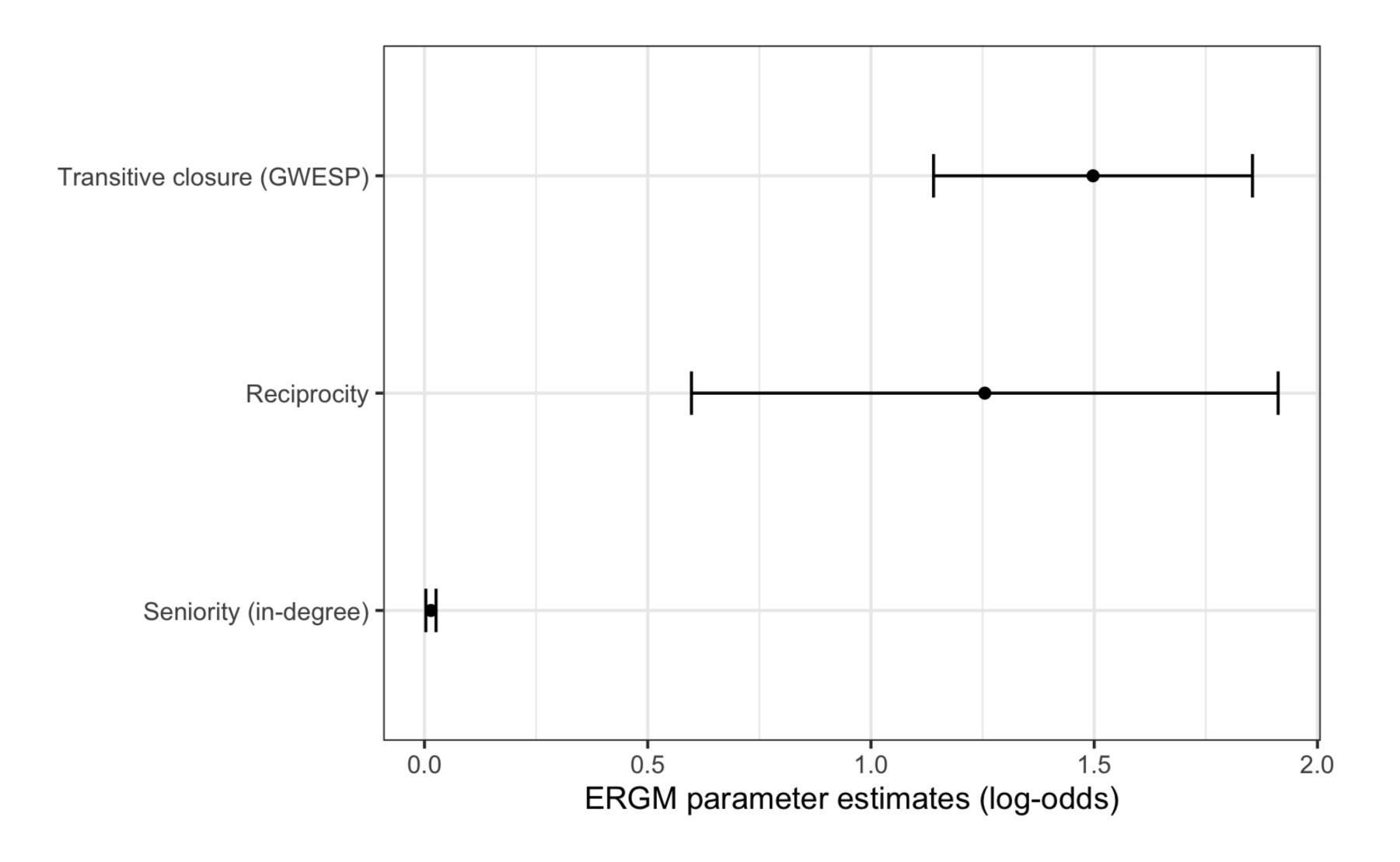
## Data

- **Data collection**: 2016 face-to-face questionnaire administration
- **Context**: freelance workers sharing a coworking space in Brescia, Italy (no shared collective identity, frequent business collaborations -> see Bianchi et al., 2018)
- Advice giving: Who do you usually turn to for advice? (Reversed edges)
- Individual attributes: seniority
- # individuals (nodes) = 29
- # ties = 120
- density = 0.15
- avg. degree = 4.10 (SD = 3.57)
- avg. seniority (months) = 29.34 (SD 14.26)





## **Evidence of reciprocation - what mechanism?**





## Instrumental framing

#### IF

costs (i.e., # of currently helped people) do not exceed a certain individual threshold

 $c_{i,t} \leq \tau_i, \quad \tau_i = \max \text{ outdegree}_i$ 

#### AND

theory: Coleman, 1991)

$$j \notin B_{i,t}$$

#### THEN

$$\rightarrow x_{ij} = 1$$





**High salience of costs**: *Ego* will help (costly transfer of resources) alter ( $x_{ii} = 1$ ) if perceived

**Conditional cooperation**: *Ego* does not help an *alter* who belongs to *ego*'s "black books" (i.e., alter has refused to help ego in the past) (shadow of the future: Axelrod, 1984; credit slip

## Solidaristic framing

IF

costs (i.e., # of currently helped people) do not exceed a certain individual threshold  $c_{i,t} \leq s_i \cdot \tau_i, \quad \tau_i = \max \text{ outdegree}_i$ AND

theory: Coleman, 1991)

 $j \notin B_{i,t}$ 

THEN

$$\rightarrow x_{ij} = 1$$

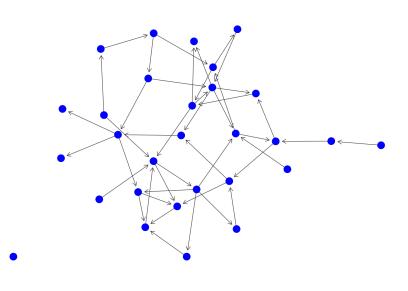


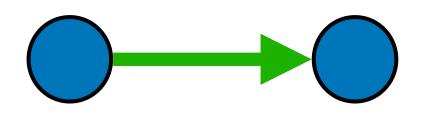
Low salience of costs: Ego will help (costly transfer of resources) alter ( $x_{ii} = 1$ ) if perceived

Sanction of opportunism: Ego does not help an alter who belongs to ego's "black books" (i.e., alter has refused to help ego in the past) (shadow of the future: Axelrod, 1984; credit slip

## Frame switch cycles

#### **Density < threshold**

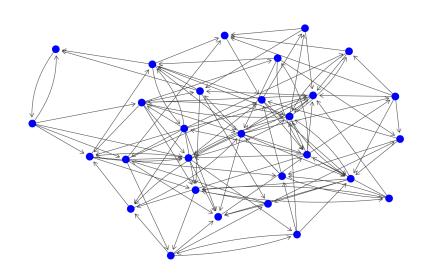


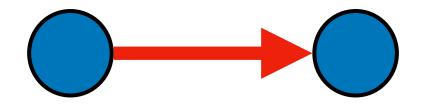


Instrumental



### **Density > threshold**

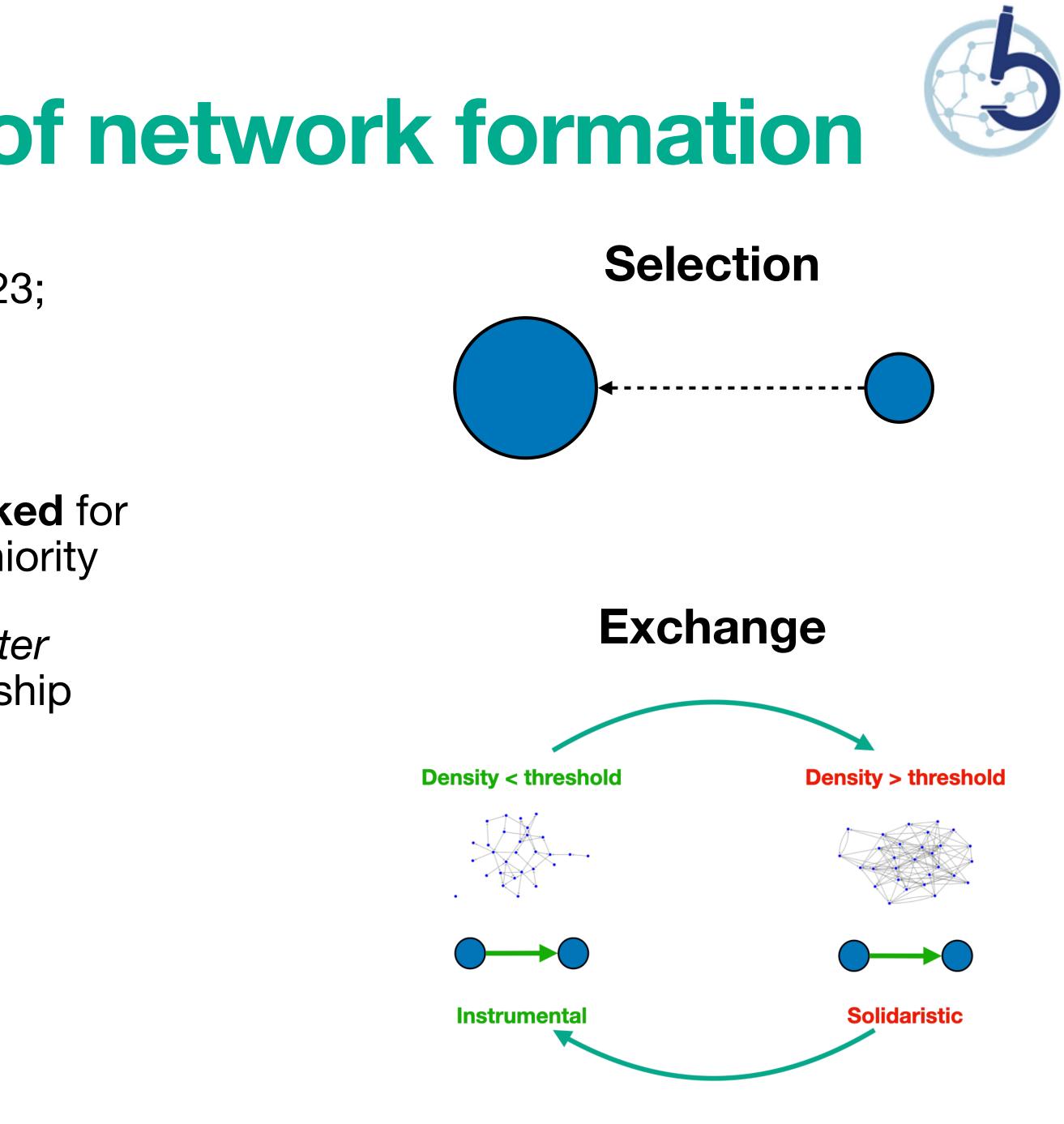




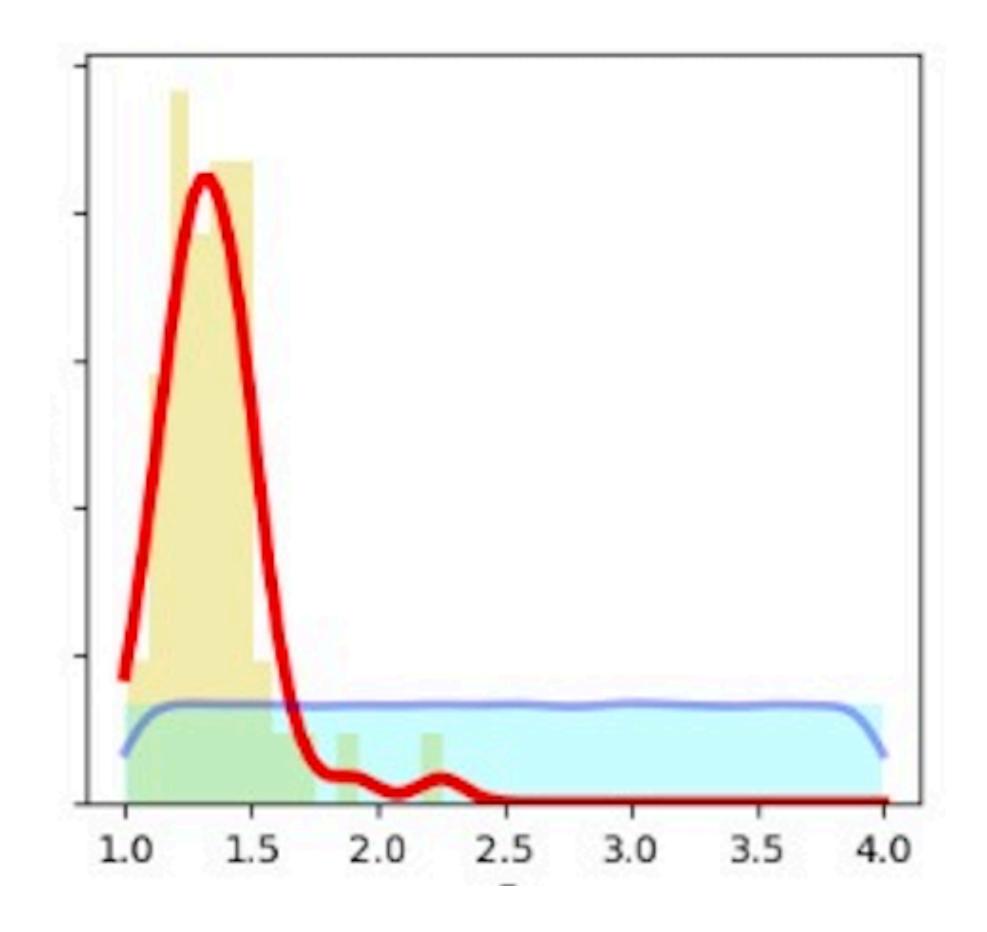
### Solidaristic

## **Agent-based model of network formation**

- ABM of the network formation (Bianchi, 2023; Bianchi & Renzini, *forthcoming*)
- **Model** of coworkers' advice exchange:
  - Selection: ego's probability of being asked for advice by *alter* as a function of ego's seniority
  - Exchange: ego sends an advice tie to alter according to their framing of the relationship
- **Estimating**:
  - Likelihood of frame switching
  - Density threshold for frame switching
- Fitting: Set of summary statistics

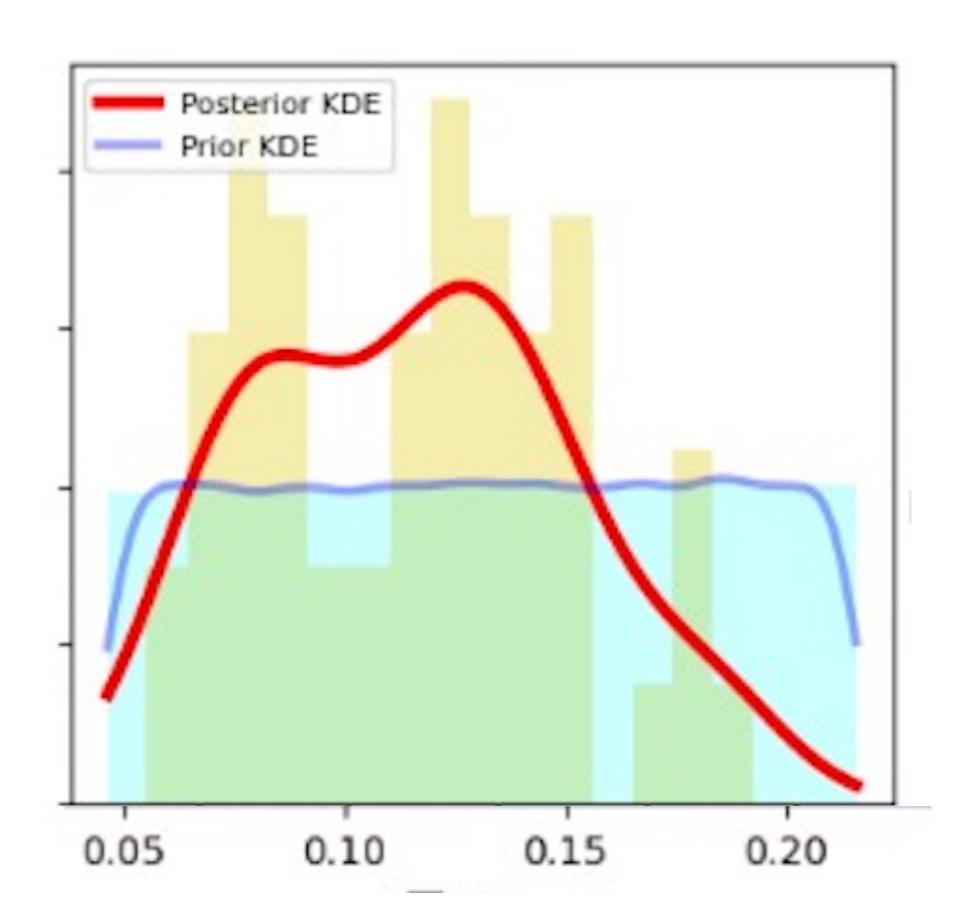


# Results: prior vs. posterior parameter distributions



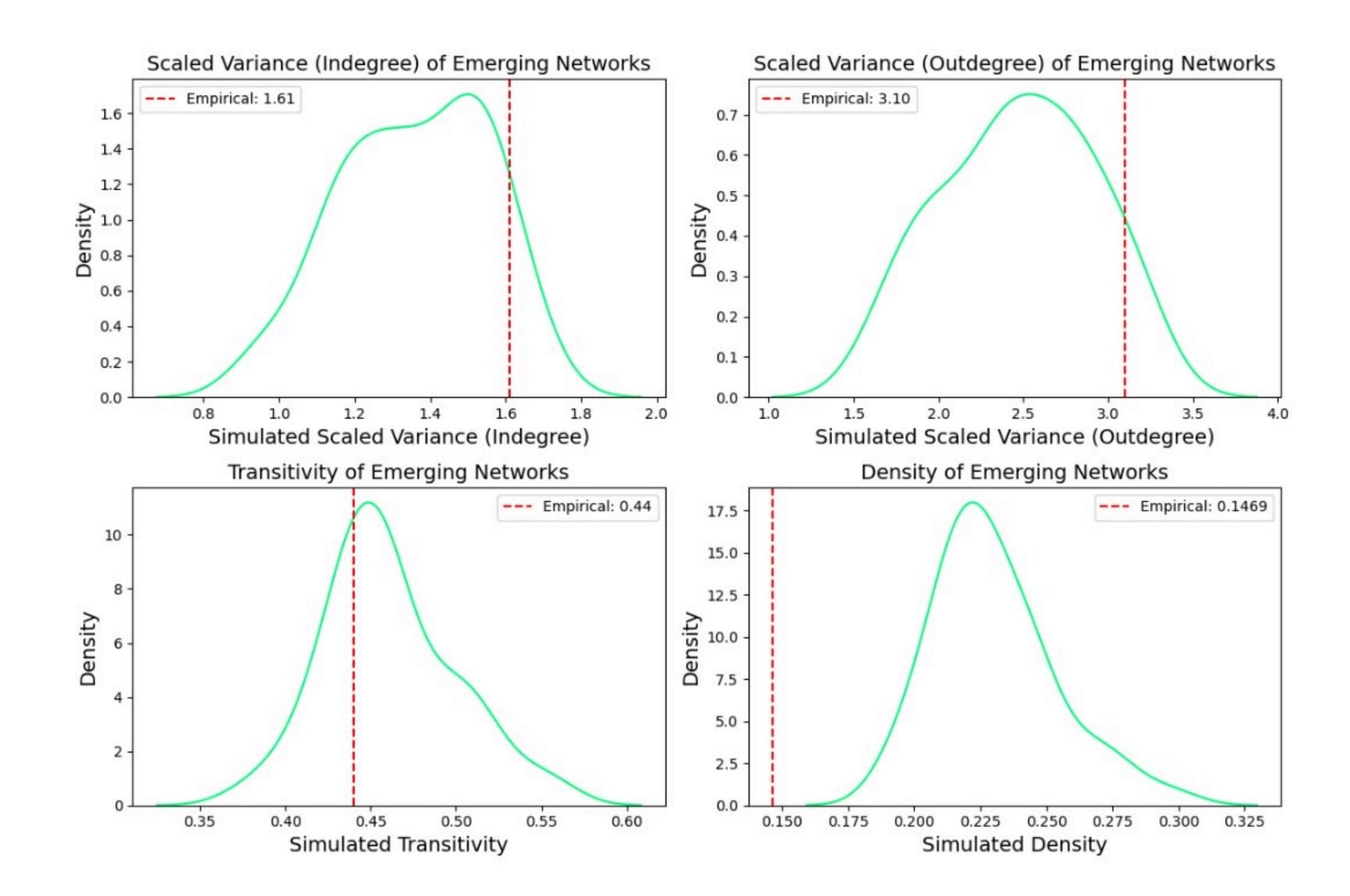
#### Cost threshold





Density threshold for frame switching

## **Model fit**





## **Estimation method**

**Approximate Bayesian Computation** (Hartig et al., 2011)

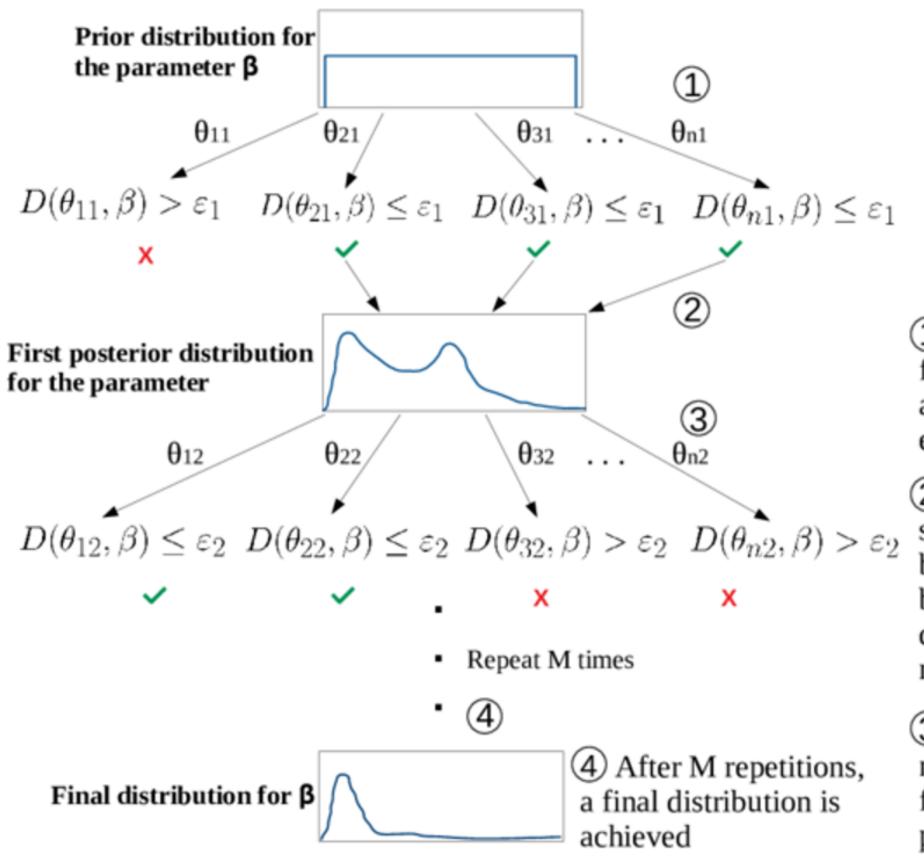
Weakly informative priors (tested with predictive checks)

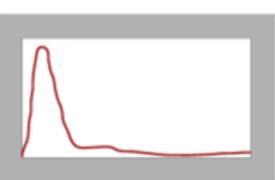
- Baseline: uniform [-3, 0]
- Threshold: {2, 3, 4, 5}
- Positive influence: uniform [0, 2.5]
- Negative influence: uniform [-2, 0]

 $D(\theta_{11},\beta) > \varepsilon_1$ 

for the parameter







True distribution for the parameter  $\beta$ 

(1) n samples  $\theta$  are randomly selected from the prior distribution and assumed as possible values for  $\beta$ . For each  $\theta$ , a simulation is performed

(2) From the n samples, those which show an error  $D(\theta_{i1}, \beta)$  in the adjusment below or equal to the tolerance  $\varepsilon_1$ become part of the posterior distribution, which is expected to be more accurate than the prior

(3) A new tolerance  $\varepsilon_2$  is placed and n samples are randomly selected from the first posterior, with a small perturbation kernel



## **Discussion points**

- Assume more complex selection processes (based on other node attributes, e.g. gender) to improve fitness on density and clustering
- Compare results to ERGM and stationary SAOM
- Fitting at observed density (simulation stopping condition) instead of equilibrium —> is it even safe to assume that we observed network in equilibrium states? ERGM needs it but bayesian estimates of an ABM doesnt't
- Cognition matters! Mechanism models ignoring context-dependent motives
  underlying behaviour might fail to adequately explain cooperation
- Empirical agent-based models can estimate the likelihood of (unobserved) cognitive components of social mechanisms



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