

Can Farmers Create Efficient Information Networks?

Experimental Evidence from Rural India

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June 2015



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(Banerjee et al. 2013, Berg et al. 2015)

The structure of networks can limit diffusion.
(Centola 2011, Golub and Jackson 2012)

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When links are chosen unilaterally, network structures that guarantee full diffusion are achieved after repeated play.
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We play this game with a population for which we have evidence of limited information diffusion: farmers in a developing country.

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- ② Does group categorisation affect network structure and efficiency?

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We contribute to the study of **information diffusion in networks**

- Foster Rosenzweig 1996, Munshi 2004, Bandiera Rasul 2006, Conley Udry 2010, Banerjee et al 2012, Ben Yishay Mobarak 2012, Berg et al 2013

We provide new evidence on how rules of thumb,
and social norms and preferences affect network structure

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Outline

1 Design

2 Predictions

3 Results

4 Conclusion

Players unilaterally create a one-way-flow network

- Subjects are divided into groups of 6
- They form links with others in the group
 - Game is divided in 2 rounds
 - In each round, every player plays once, in random, unknown order
 - Only 1 link allowed. In round 2 players rewire old link
- One player is randomly drawn to win a monetary prize
- Players directly or indirectly connected in the final network to the winner also win the prize

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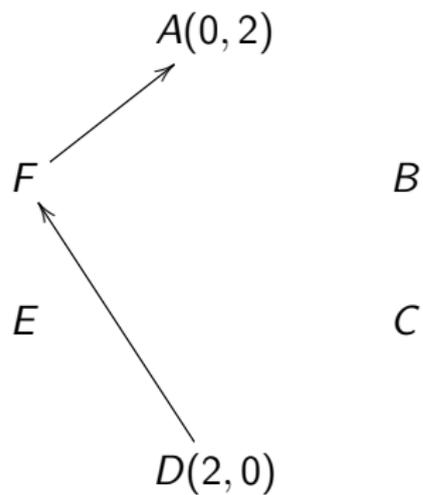
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Reach, in-reach



What will an efficient information network look like?

A

F

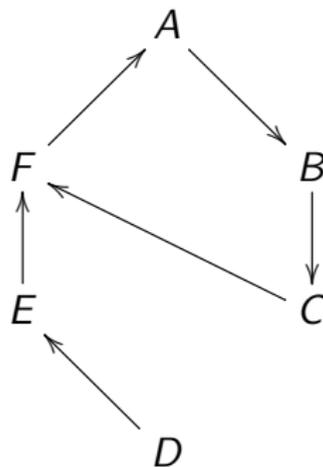
B

E

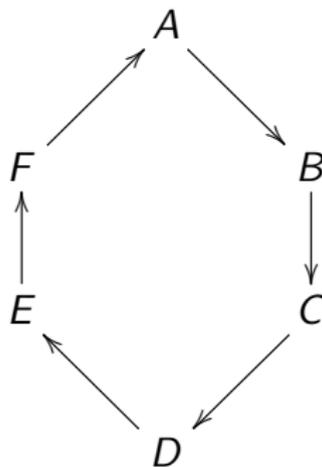
C

D

An inefficient network



The efficient cycle network



A simple measure of efficiency

$$\text{Efficiency}_g = \frac{\frac{1}{n} \sum_{i=1}^n \text{reach}_i}{5} \quad (1)$$

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In Treatment 1 players observe the information of others

- Each player i can form a link towards a player j in the group
- Myopic, selfish best response: select the person with the most information, ie. the **maximum reach**
- If every player myopically best responds, the network configuration converges towards the cycle
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An example

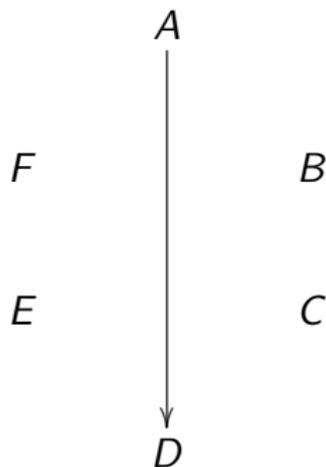


Figure : Move 1

An example

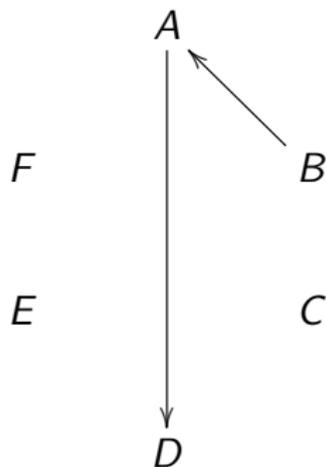


Figure : Move 2

An example

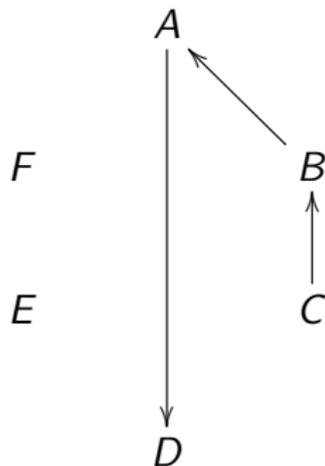


Figure : Move 3

An example

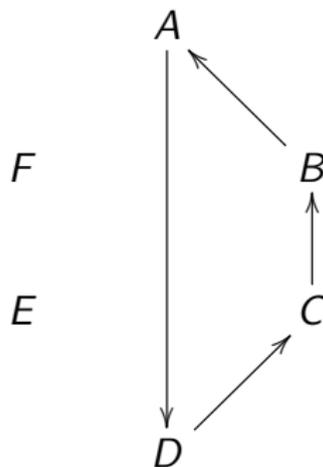


Figure : Move 4

An example

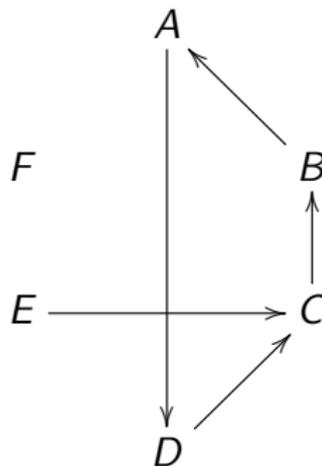


Figure : Move 5

An example

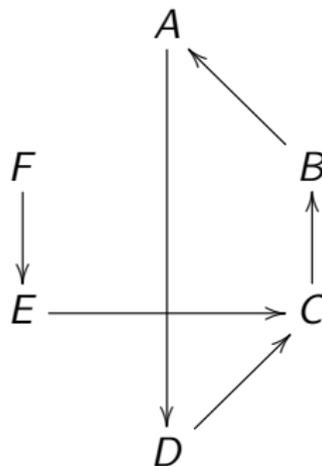


Figure : Move 6

An example

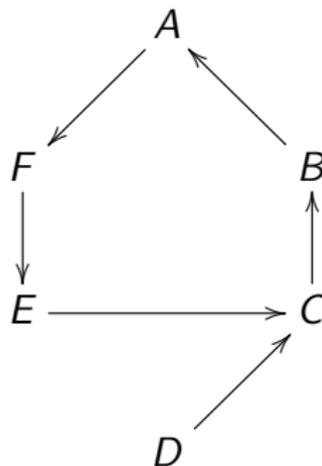


Figure : Move 7

The cycle!

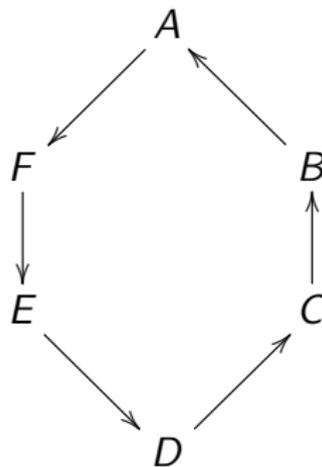


Figure : Move 11

In T2 players share own information with others

- Each player i can form a link from j to i
- If social welfare maximiser: pick player who passes prize to the largest number of people, ie **maximum in-reach**
 - Network converges towards the cycle in (almost) all cases
- If Rawlsian: **minimum reach**

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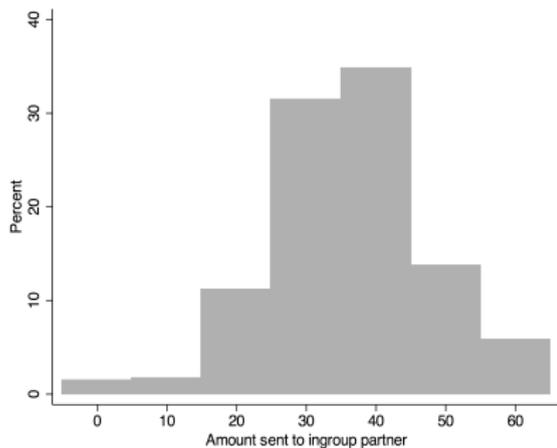
The effects of group categorisation

- We impose exogenous social identities, create saliency, and make them public in half of the sessions
- Akerlof Kranton (2000) suggest individuals get positive utility from following group prescriptions

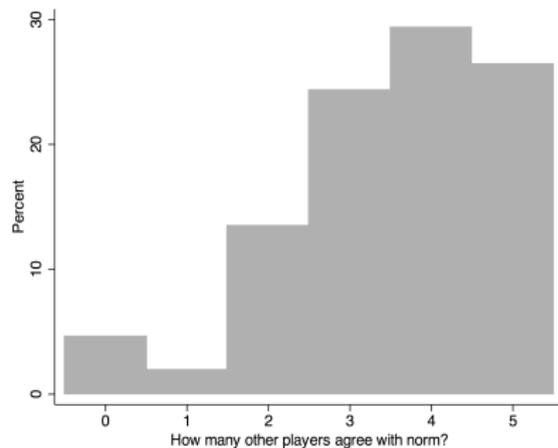
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Group salience and in-group norms



(a) Monetary allocation



(b) Normative expectation

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The experiment

- Villages from 1.30 to 3hr from Pune, Maharashtra
 - Random door-to-door sampling of adult male farmers
- 486 participants in 81 sessions
- Balance on the available set of covariates
- 80 percent made at most 2 mistake out of the 7/8 understanding questions

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- We use non-parametric **Wilcoxon rank sum tests** for session level outcomes
- and dyadic regression analysis for individual data:

$$\text{link}_{ij,r} = \alpha + \beta \mathbf{Network Position}_{j,r} + \gamma \mathbf{D}_{ij} + \delta \text{round}_r + e_{ij,r}$$

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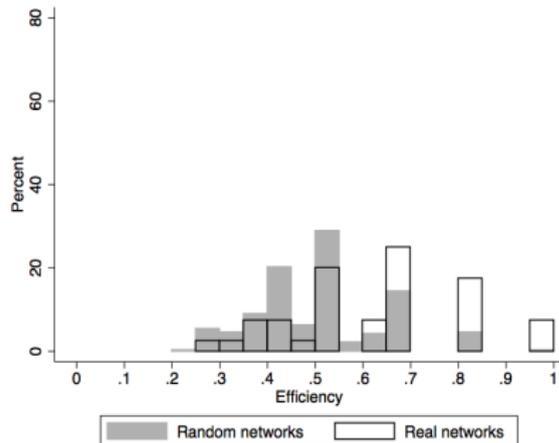
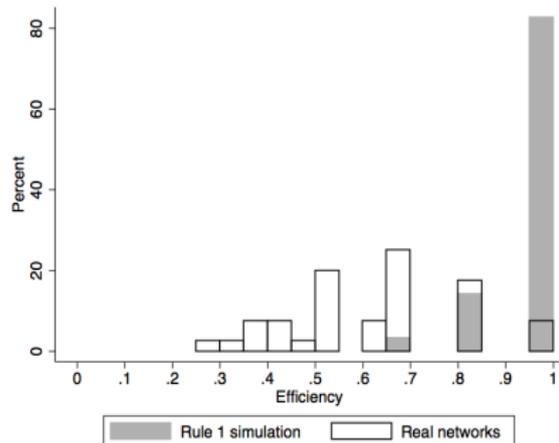
$$\text{link}_{ij,r} = \alpha + \beta \mathbf{Network Position}_{j,r} + \gamma \mathbf{D}_{ij} + \delta \text{round}_r + e_{ij,r}$$

Figure : A session of the experiment



Result 1: Efficiency is 64 percent, significantly lower than what the simple rules would achieve

Figure : Efficiency in no-identity sessions and in simulated networks



(a) Rank sum test: $Z = 12.08$, $p < .001$

(b) Rank sum test: $Z = 4.62$, $p < .001$

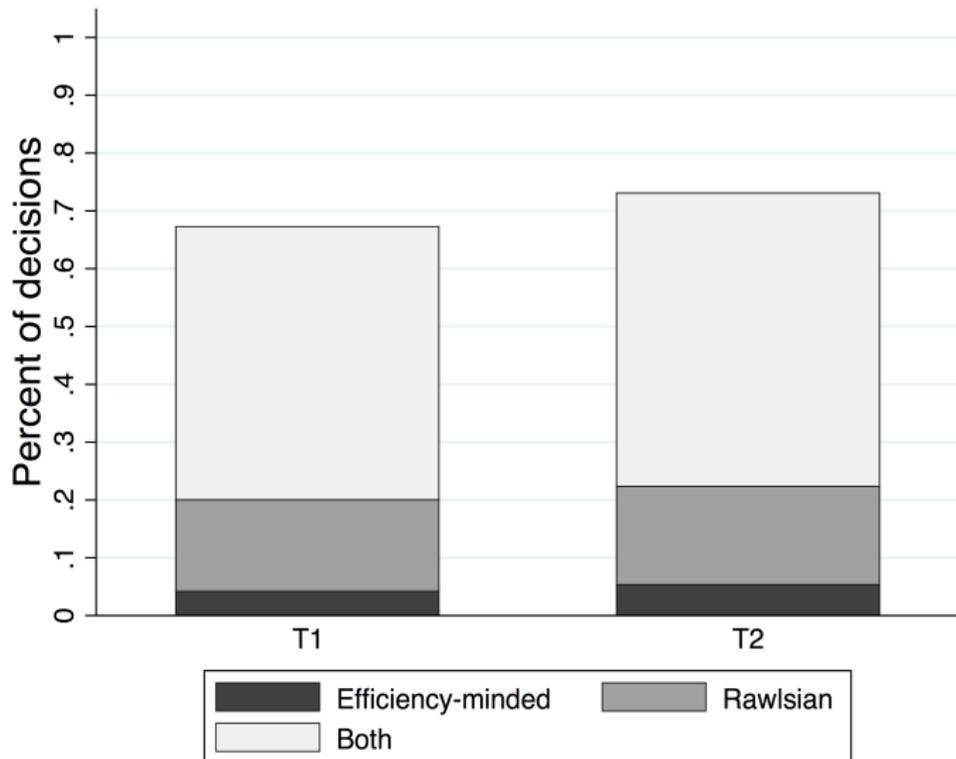
Result 2: Reach and in-reach predict new links

Table : Dyadic linear probability model

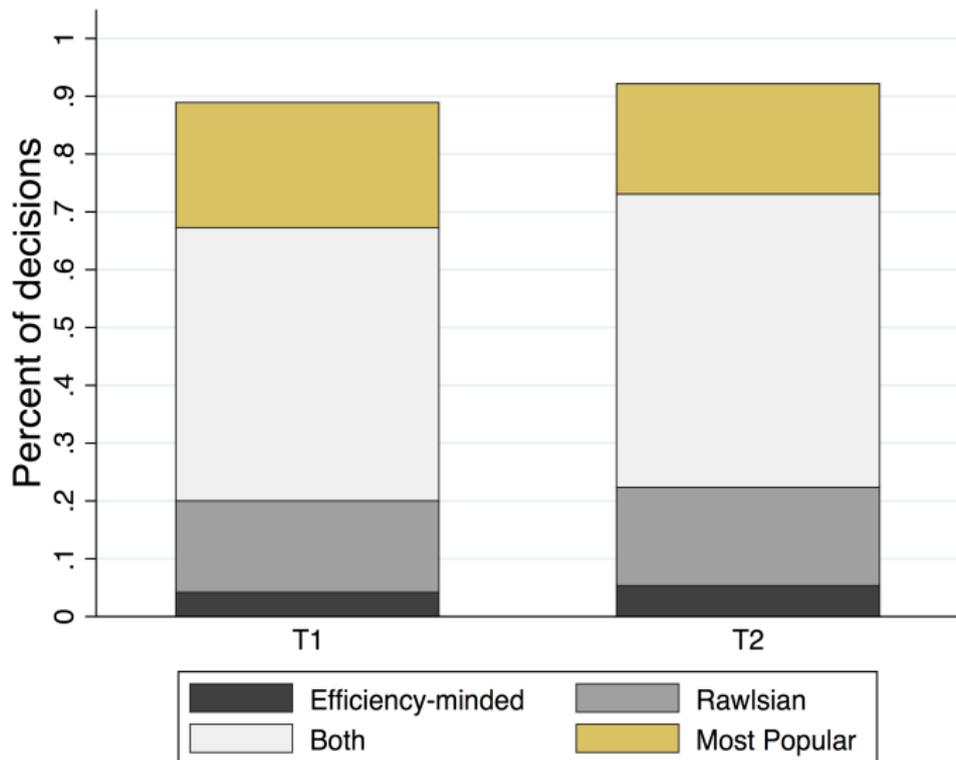
	(1)	(2)	(3)	(4)
Panel a				
max reach _j	.132 (.001) ^{***}	.130 (.001) ^{***}		
min in-reach _j	.018 (.235)	.016 (.314)		
max in-reach _j			.111 (.004) ^{***}	.120 (.002) ^{***}
min reach _j			.073 (.04) ^{**}	.066 (.072) [*]
Sessions	T1	T1	T2	T2
Obs.	1200	910	1260	940
Cluster N	20	20	21	21
Controls		V		V

Dyadic OLS regression. Dependent variable is a dummy which takes a value of one if *i* chose to establish a link with *j*. Each regression contains controls for a set of observables, round dummies and dummies for each possible pairing of map positions. Confidence: *** ↔ 99%, ** ↔ 95%, * ↔ 90%. Standard errors are corrected for clustering at session level. P-values obtained with wild bootstrap-t procedure reported in parentheses.

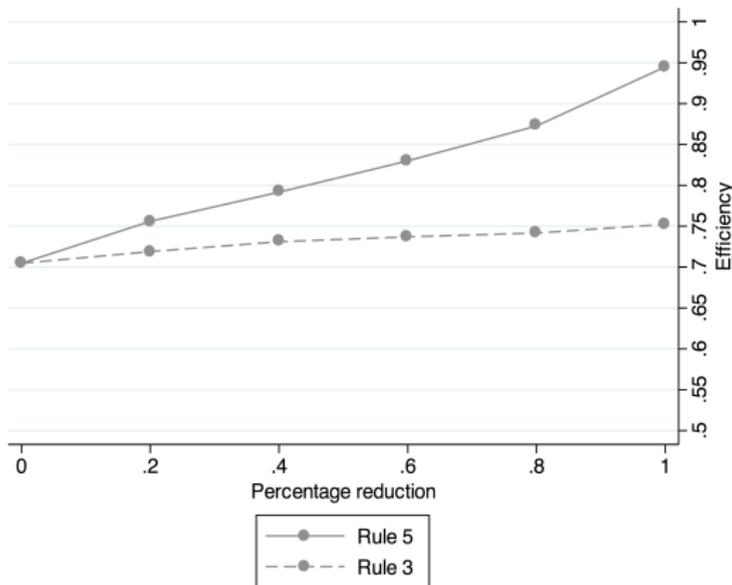
Links to the most popular player



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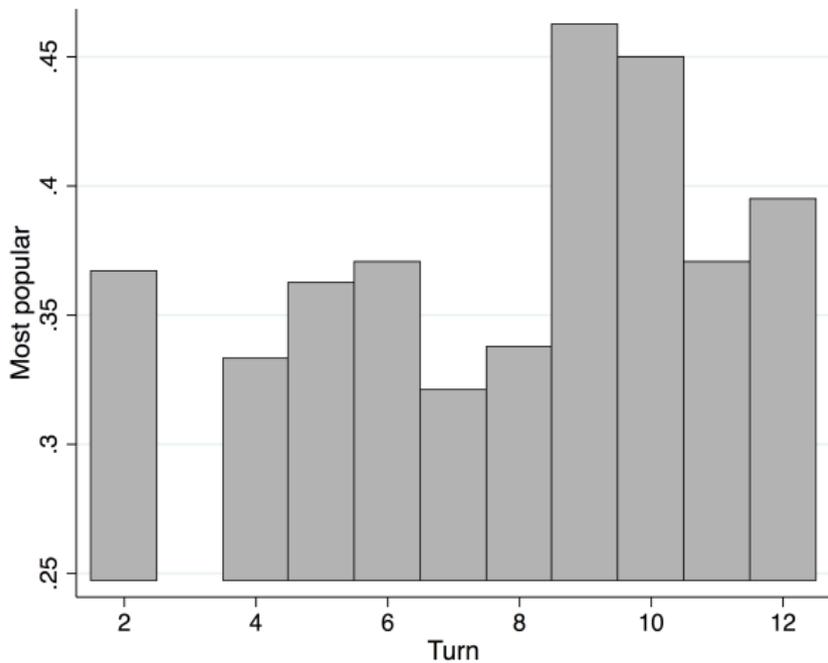


Result 3: Rule 5 causes the highest efficiency losses



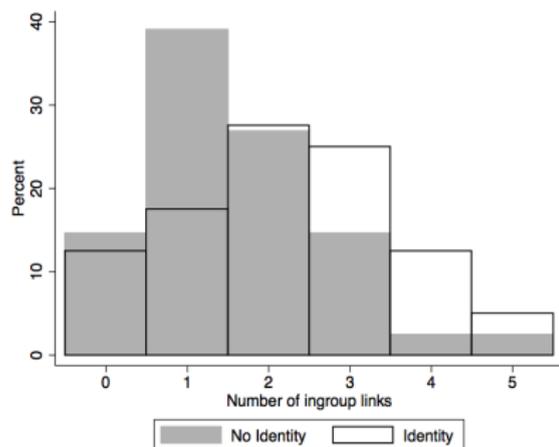
Note. In the baseline simulation 54 percent of decisions follow rule 2, 16 percent follow rule 3, and 30 percent follow rule 5. Each point in the graph represents average efficiency over 100 repetitions of the link formation game.

Figure : Proportion of links that target the 'most popular' player

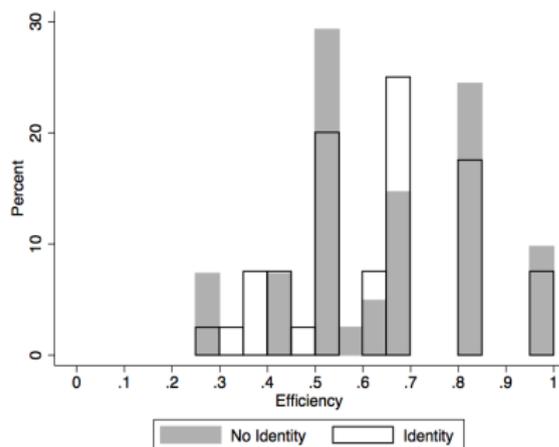


Result 4: Ingroup links significantly increase, but there is no reduction in efficiency

Figure : Identity and no-identity sessions



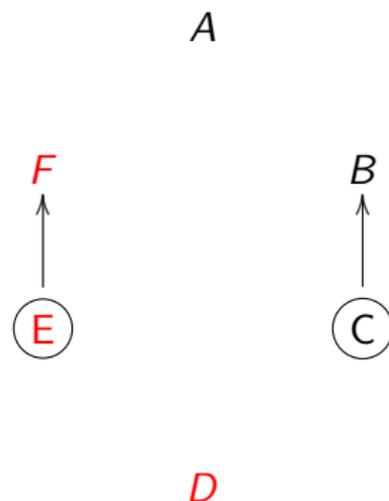
(a) Ingroup links
 $Z = 2.23, p = .02$



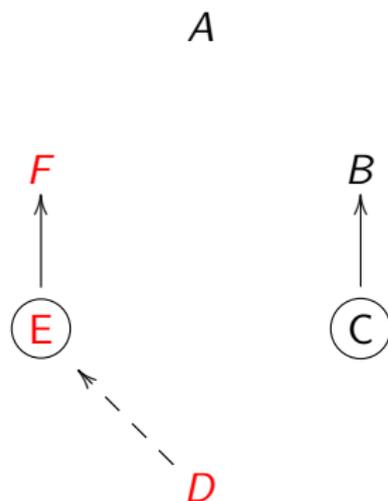
(b) Efficiency
 $Z = -0.51, p = .61$

How does this come about?

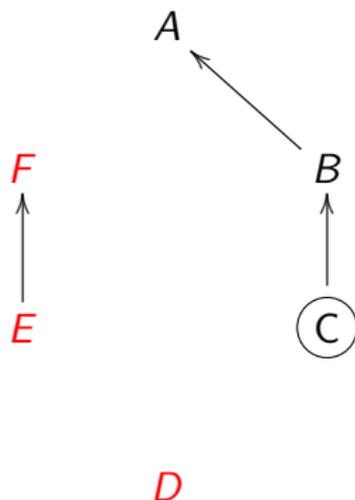
(i) Best response set includes both in and out group players



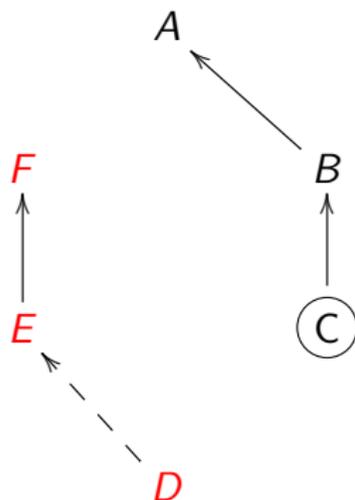
Ingroup links \uparrow , efficiency =



(ii) Best response set includes only out group players



Ingroup links \uparrow , efficiency \downarrow



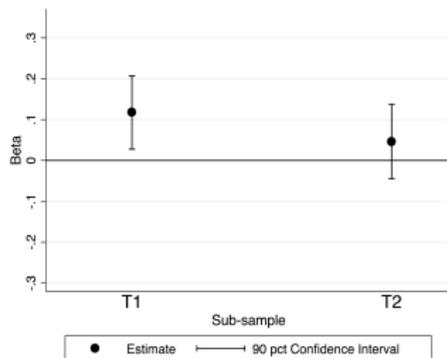
If group concerns are weak (these are arbitrary groups!), then farmers will choose in-group links only in case (i)

We will use the following model to explore mechanisms further:

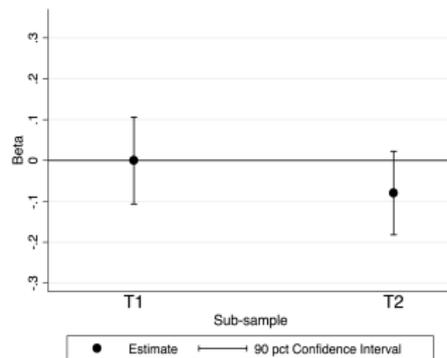
$$x_{dis} = \alpha + \beta_1 \text{Identity Session}_s + e_{dis} \quad (2)$$

In-group links increase, efficiency minded links do not

Figure : Linear probability model (2): coefficient estimates

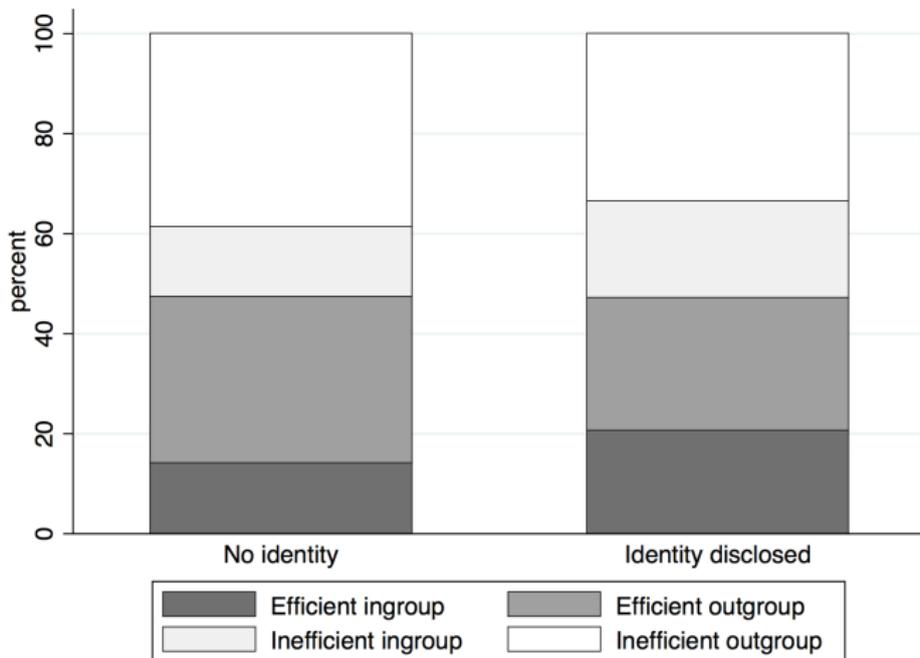


(a) x_{dis} =: in-group link



(b) x_{dis} = efficiency-minded link

Efficiency minded in-group links increase



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Network efficiency is 32 percentage points below potential.

⇒ Future work needs to validate this with real networks and document the extent of belief convergence in communities

We show that group membership and a concern for other farmers' welfare affect link formation.

⇒ Important to understand information flows in society and the optimal design of peer-to-peer interventions

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Balance

Table : Balance test: Identity Sessions

	Age	Edu	UpperCaste	LandOwned	LandCult	NetSize
	(1)	(2)	(3)	(4)	(5)	(6)
Identity	-.194 (1.764)	.029 (.056)	-.087 (.067)	.063 (.517)	.101 (.468)	-.201 (1.100)
Obs.	479	466	433	475	470	428

OLS regressions. The dependent variable is indicated in the row's name. Upper caste is a variable that takes value of 1 if respondent is not from a schedule caste, a scheduled tribe or an Other Backward Caste. Network size is the self reported number of peers with whom the farmer exchanges advice on agricultural matters. Confidence: *** \leftrightarrow 99%, ** \leftrightarrow 95%, * \leftrightarrow 90%. Standard errors clustered at the session level reported in parentheses.

Balance

Table : Balance test: T2 sessions

	Age	Edu	UpperCaste	LandOwned	LandCult	NetSize
	(1)	(2)	(3)	(4)	(5)	(6)
T2	-1.582 (1.761)	-.028 (.056)	-.052 (.068)	-.085 (.514)	-.049 (.465)	1.293 (1.089)
Obs.	479	466	433	475	470	428

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Balance

Table : Balance test: Identity in T1 sessions

	Age	Edu	UpperCaste	LandOwned	LandCult	NetSize	Und
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Identity	-2.378 (2.544)	.091 (.081)	-.040 (.098)	.077 (.737)	.141 (.657)	.111 (1.102)	-.267 (.178)
Obs.	235	232	215	234	231	211	240

OLS regressions. The dependent variable is indicated in the row's name. Upper caste is a variable that takes value of 1 if respondent is not from a schedule caste, a scheduled tribe or an Other Backward Caste. Network size is the self reported number of peers with whom the farmer exchanges advice on agricultural matters. Und is the number of mistakes in the initial 7 understanding questions. Confidence: *** \leftrightarrow 99%, ** \leftrightarrow 95%, * \leftrightarrow 90%. Standard errors clustered at the session level reported in parentheses.

Balance

Table : Balance test: Identity in T2 sessions

	Age	Edu	UpperCaste	LandOwned	LandCult	NetSize	Und
	(1)	(2)	(3)	(4)	(5)	(6)	(7)
Identity	1.879 (2.400)	-.033 (.076)	-.135 (.093)	.046 (.733)	.061 (.673)	-.482 (1.877)	-.224 (.197)
Obs.	244	234	218	241	239	217	246

OLS regressions. The dependent variable is indicated in the row's name. Upper caste is a variable that takes value of 1 if respondent is not from a schedule caste, a scheduled tribe or an Other Backward Caste. Network size is the self reported number of peers with whom the farmer exchanges advice on agricultural matters. Und is the number of mistakes in the initial 8 understanding questions. Confidence: *** \leftrightarrow 99%, ** \leftrightarrow 95%, * \leftrightarrow 90%. Standard errors clustered at the session level reported in parentheses.