







THE HAGUE

Network Science for Official Statistics **Anonymity and Disclosure Control for Network Data**

Rachel de Jong, Mark van der Loo, Frank Take Leiden university, CBS October 5th 2025

Towards Anonymity in Network Data

Microdata

ID

1

2

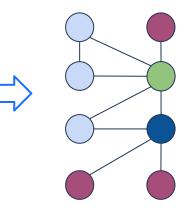
3

4

Attribute

5 6 7 8

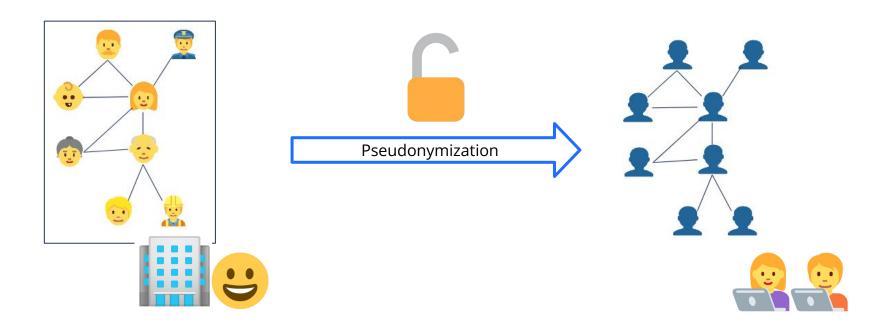
Network data



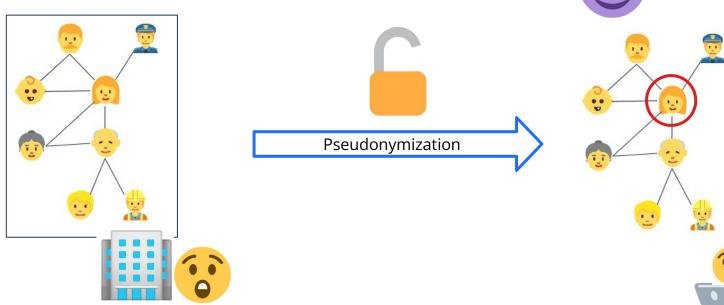
Network consists of

- Nodes 0
 - ☐ Entities: people, companies, cities, ...
- Edges
 - **Connections**: social ties, trade, roads, ...

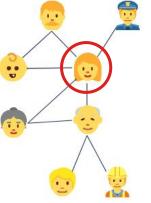
Safely publishing networks?



Safely publishing networks?

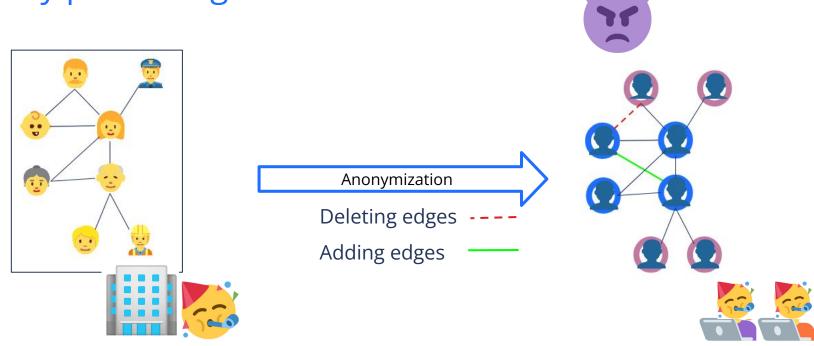








Safely publishing networks?



Q1: How to **measure** anonymity?

- *k*-Anonymity
- Measure = attacker scenario
- Uncertain information?

Q2: How to **anonymize** a network?

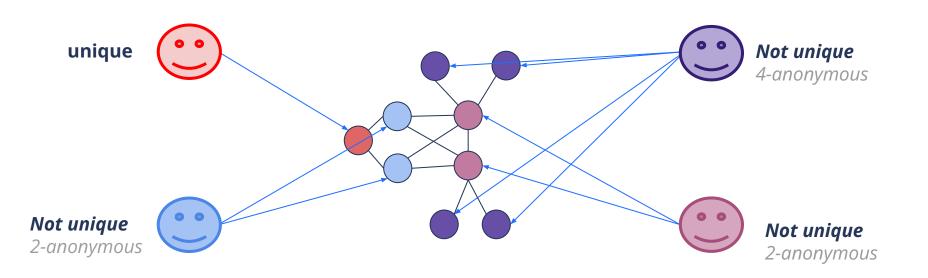


k-Anonymity

How many candidates are there for each individual?

For now: **Not unique** → **Anonymous**

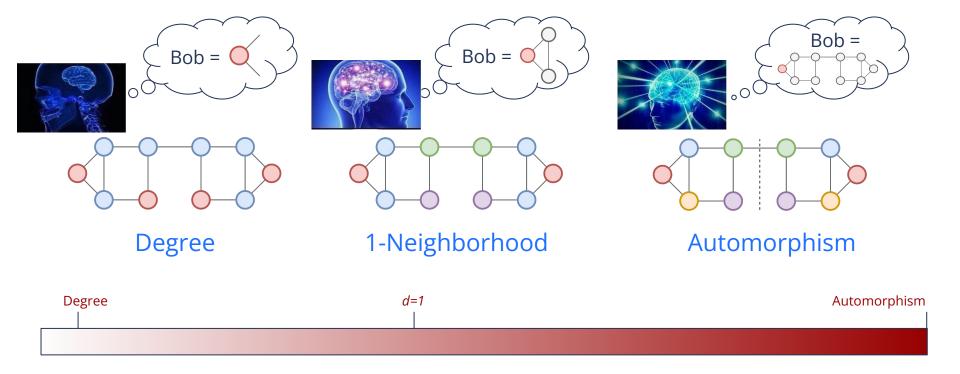
anonymity = uniqueness: fraction of unique nodes



K-anonymity Not unique

 \longrightarrow

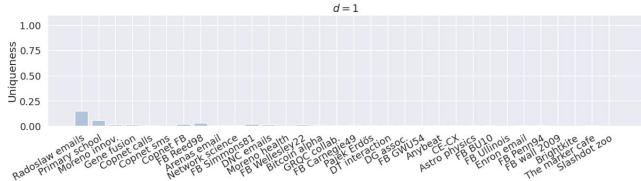
Anonymous*





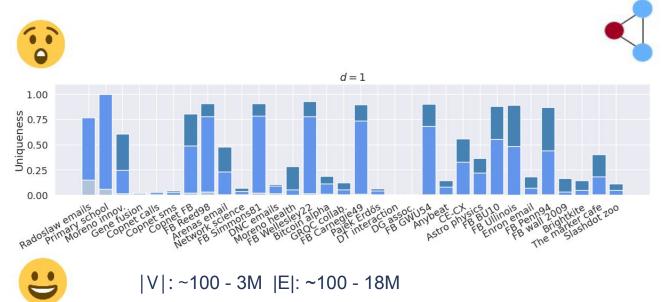


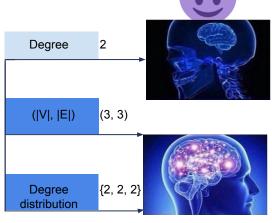
Degree

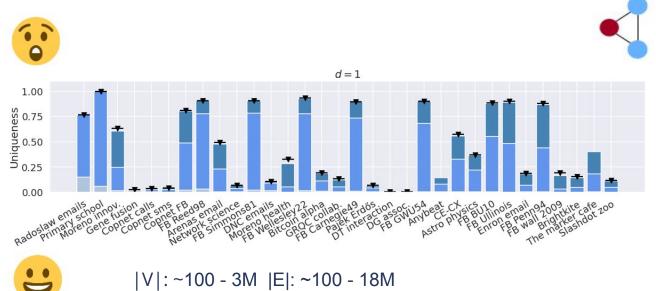


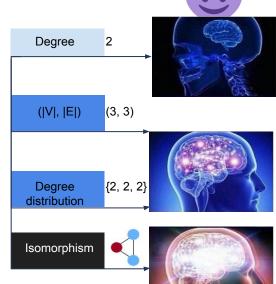


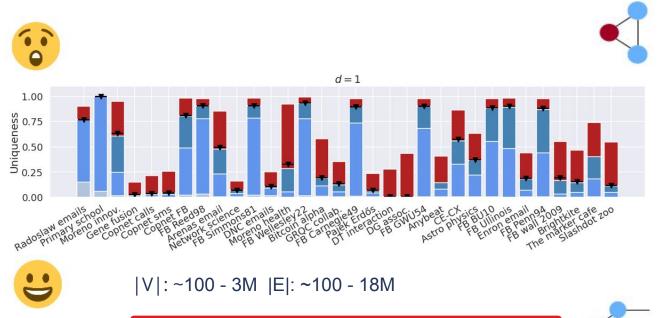
|V|:~100 - 3M |E|:~100 - 18M









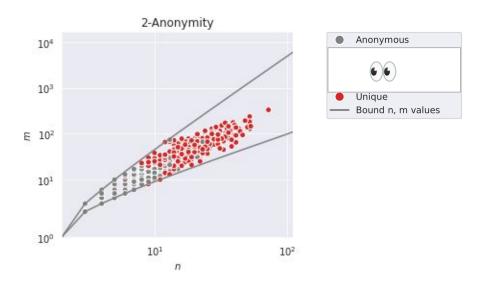


Degree distribution {2, 2, 2} distribution {2, 2, 2} Degrees of neighbors {2, 3, 3}

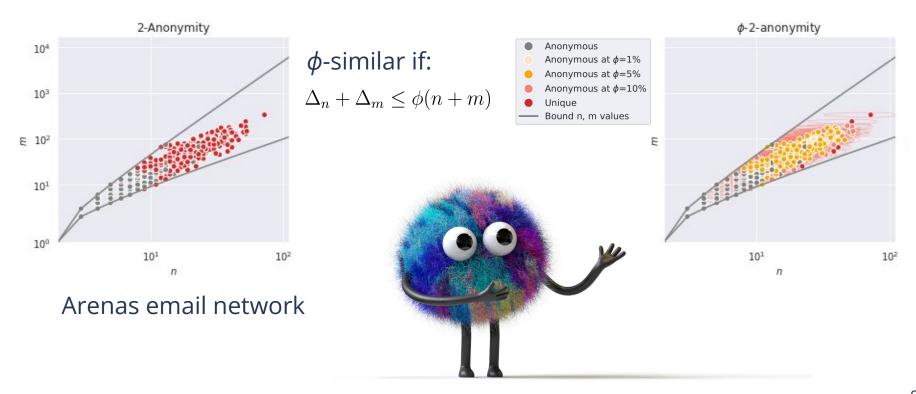
Degree

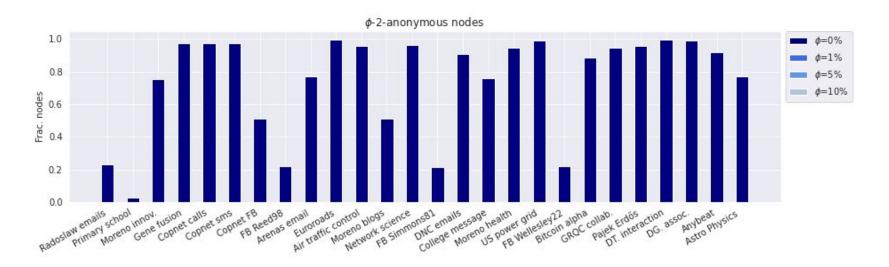
Imprecise information but looking further is more effective!

What if the attacker is not so sure?

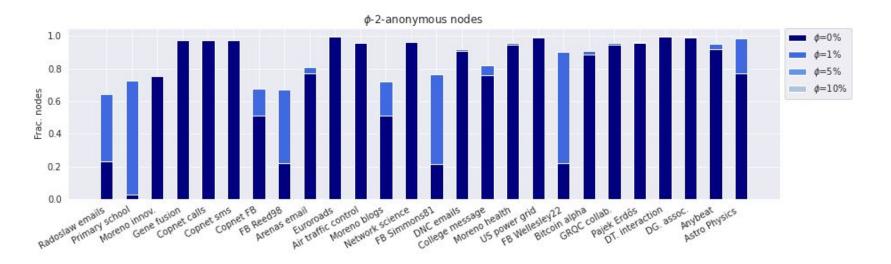


Arenas email network



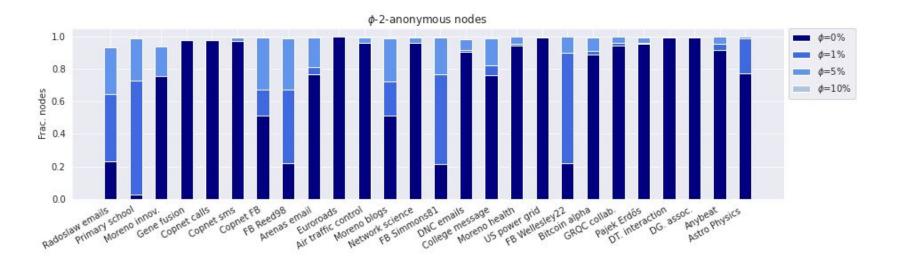


No uncertainty: >2% anonymous for all networks



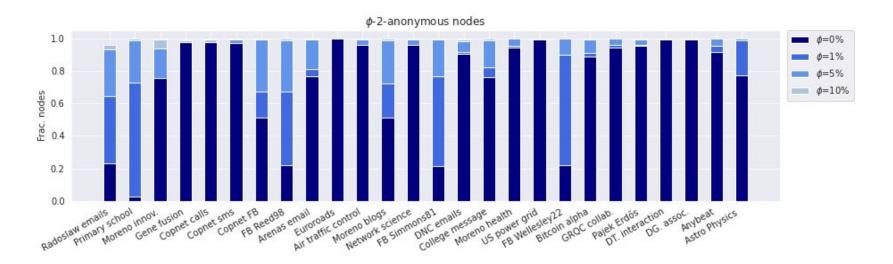
No uncertainty: >2% anonymous for all networks

1% uncertainty: >60% anonymous



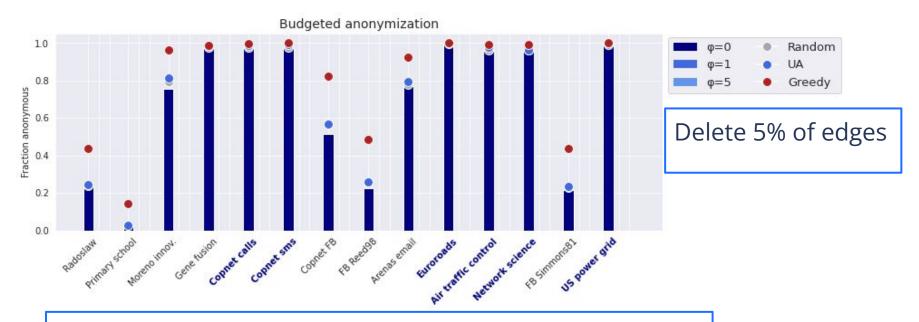
No uncertainty: >2% anonymous for all networks

1% uncertainty: **>60%** anonymous 5% uncertainty: **>90%** anonymous

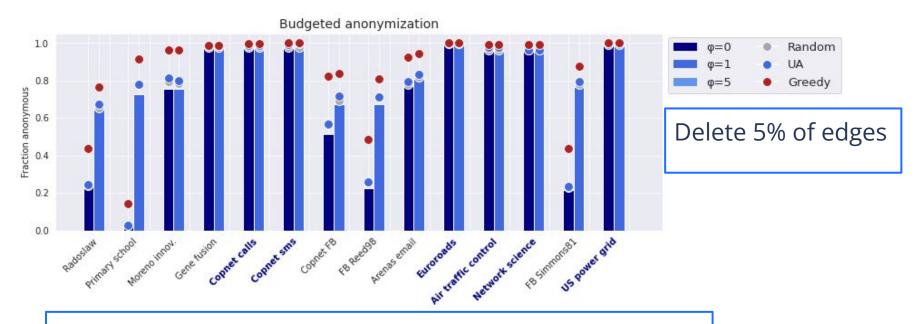


No uncertainty: >2% anonymous for all networks

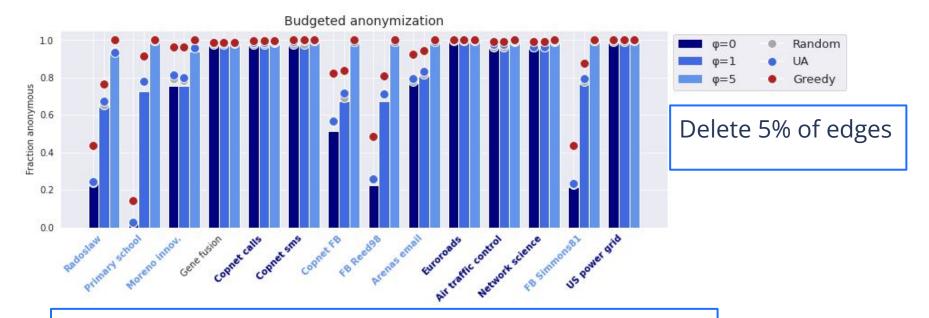
1% uncertainty: **>60%** anonymous 5% uncertainty: **>90%** anonymous 10% uncertainty: **>95%** anonymous



0%: anonymization limited effect



0%: anonymization limited effect



0%: anonymization limited effect

5%: most networks anonymous



Conclusion

- Measure anonymity with *k*-anonymity
- Attacker knowledge = anonymity measure
- Imprecise attacker knowledge
 - → large increase in anonymity
- Anonymization + 5% uncertainty
 - →Many networks fully anonymous

Future work

- Discussion: which level of uncertainty is likely?
 - Guidelines ϕ values
- Fuzzyness for other measures
- Include more network properties
 - Timestamps, node / edge labels







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ANONET package

github.com/RacheldeJong/ANONET



- [1] The effect of distant connections on node anonymity in complex networks. Scientific Reports 14(1), 1156 (2024)
- [2] Algorithms for efficiently computing structural anonymity in complex networks. ACM Journal of Experimental Algorithmics 28 (2023)
- [3] A systematic comparison of measures for k-anonymity in networks (2024) arXiv:2407.02290 [4] The anonymization problem in social networks, preprint arXiv:2409.16163 (2024).

Authors: R.G. de Jong, M. P. J. van der Loo, F. W. Takes

Related research

How to **measure** anonymity?*

- Algorithms for efficiently computing structural anonymity in complex networks. ACM Journal of Experimental Algorithmics, 28, 1-22. (2023).
- The effect of distant connections on node anonymity in complex networks. Scientific Reports, 14(1), 1156. (2024)
- A systematic comparison of measures for k-anonymity in networks. arXiv preprint arXiv:2407.02290. (2024)
- Paper on fuzzy k-anonymity in progress...

How to **anonymize** a network?

The anonymization problem:

de Jong, R. G., van der Loo, M. P., & Takes, F. W. The anonymization problem in social networks. arXiv preprint arXiv:2409.16163. (2024)

Genetic algorithm for anonymization:

Bonello, S., de Jong, R. G., Bäck, T. H., & Takes, F. W.
 Utility-aware Social Network Anonymization using
 Genetic Algorithms. In Proceedings of the Genetic and Evolutionary Computation Conference Companion (pp. 775-778). (2025)

Simulated Annealing for network anonymization:

E.D. Arsene, R.G. de Jong, F.W. Takes, and A.L.D. Latour,
 A Simulated Annealing Approach to Social Network
 Anonymization (under review)

*Authors: Rachel G. de Jong, Mark P.J. van der Loo, Frank W. Takes

Appendix

Related work: Hackathon*

Which information can be obtained?

May 3rd 2022 In 4 hours:

- 22 students found > 5,000 links about 26 volunteers
- Different sources: Facebook, linkedin, instagram (images)
- Different categories: Family, work, school
- 11 large pizzas

Co-organisers: Mark van der Loo, Jayshri Murli, Marieke de Vries, Peter-Paul de Wolf and, Frank Takes

*de Vries, M. M., et al. "The risk of identity disclosure through network structure: anec-dotal evidence from a hackathon." arXiv preprint arXiv:2211.15325 (2022).

HOW ANONYMOUS ARE YOU?!

How anonymous are you in openly available data? Is it possible for hackers to extract sensitive information from these data sets, by combining them with other data sources?

In this hackathon you, as a 'potential attacker', will search for network relations of a given list of people. If you are able to find the most relations within 4 hours, you receive a prize!

MAY 3, 15:00 - 19:00



Free pizza and beverages afterwards!

SCAN TO
REGISTER

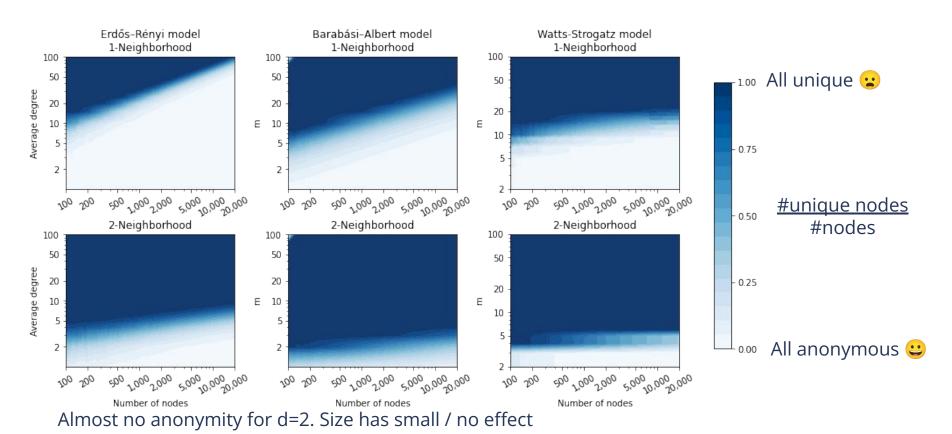
Sign up
before 27
April.



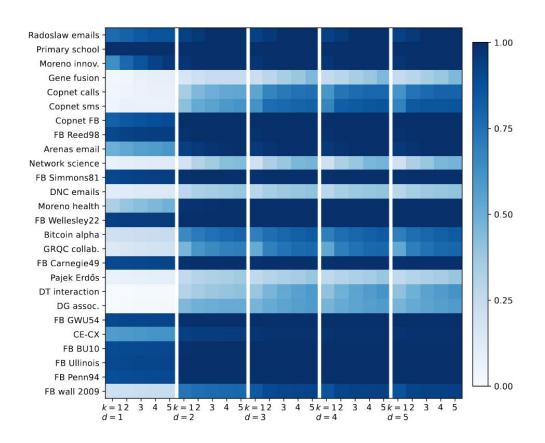




d-k-Anonymity on graph models



d-k-Anonymity: d> 2, k > 1



$$d=1 \rightarrow d=5$$

- \Box *d*=1 is not enough
- \square After d=2 often not much change

$$k=1 \rightarrow k=5$$

- ☐ Average: +0.05 up to 0.08
- ☐ For some networks larger effect

Focus on:

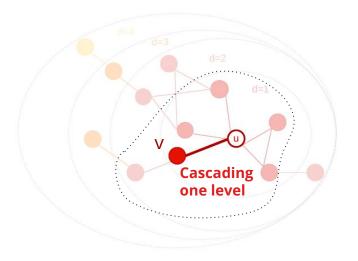
- ☐ 2-neighborhood
- ☐ Uniqueness (k=1)

Approach 2: Anonymity-cascade

Bob = v

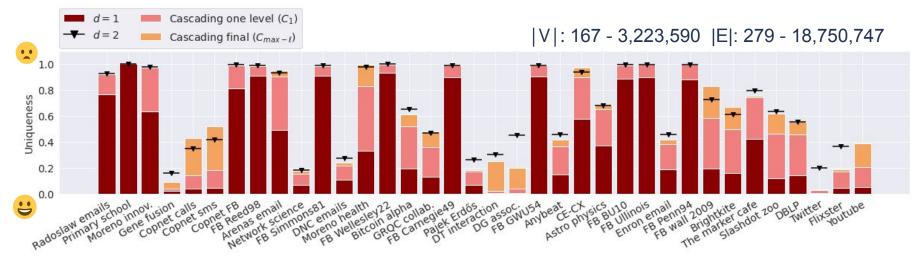
Knowledge:

- 1-neighborhood
- Connected to unique node
 - "Infecting nodes with uniqueness"





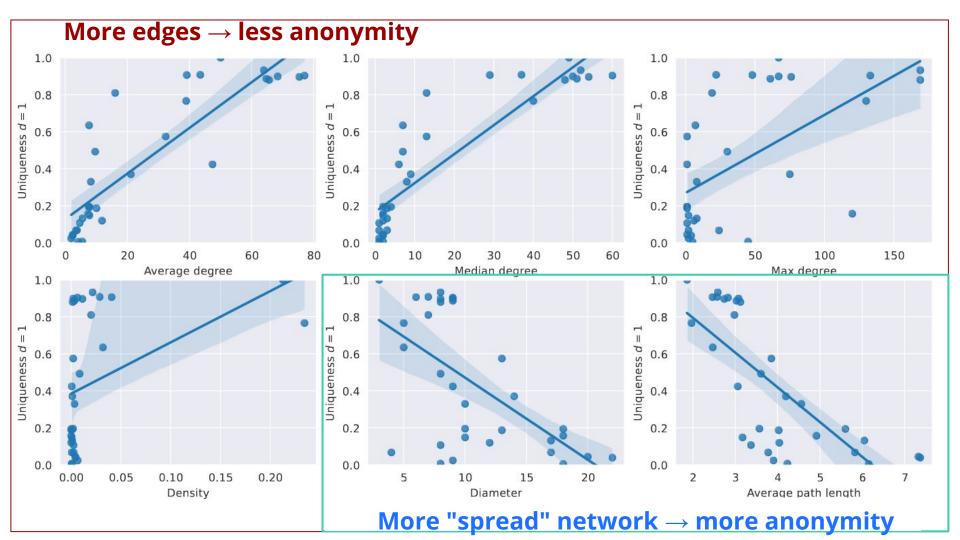
Anonymity-Cascade on real networks



*More networks due to lower runtimes d=2 and cascading

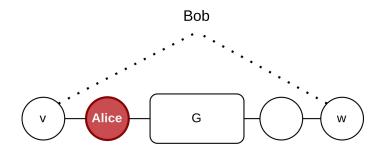
>100% increase: 19 / 5 / 36

Large increase in uniqueness with knowledge of one connection Approximate d-k-anonymity with d=2 Increase for some networks with cascading final



Limitation: Twin Nodes

So far we have assumed that k candidates are distant

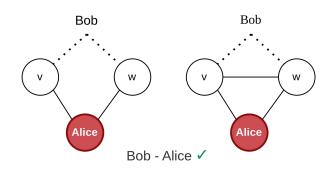


Bob - Alice ?

v and *w* structurally indistinguishable (same orbit)



What if they are *both* connected to Alice?



Open twins: $N_1(v) - \{v\} = N_1(w) - \{w\}$ Closed twins:

 $N_1(v) = N_1(w)$

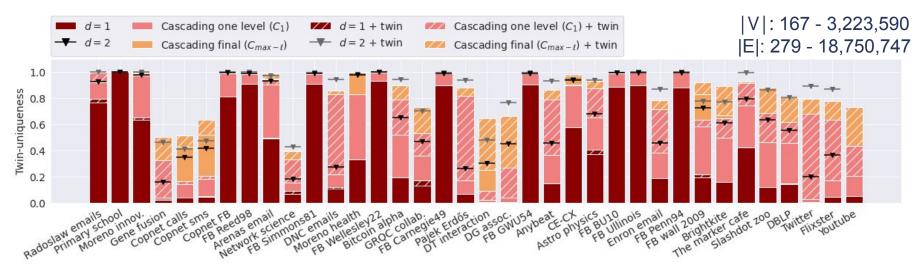
 $\rightarrow v$, w structurally indistinguishable





	IVI	IEI	Fraction twins	ℓ_{max}	d=1 (s)	d=2 (s)	C_1	(s) C_f (s)		
Radoslaw emails ¹⁹	167	6,500	0.036	3	0.02 s	< 0.01 s	< 0.01 s	< 0.01 s	Communication	
Primary school ²⁰	236	11,798	0.000	1	0.03 s	< 0.01 s	< 0.01 s	< 0.01 s	Human contact	
Moreno innov. 19	241	1,846	0.012	3	0.00 s	0.01 s	< 0.01 s	< 0.01 s	Communication	
Gene fusion ¹⁹	291	558	0.581	6	0.00 s	< 0.01 s	< 0.01 s	< 0.01 s	Bio	
Copnet calls ¹	499	4,180	0.024	4	0.00 s	0.02 s	< 0.01 s	< 0.01 s	Communication	
Copnet sms ¹	570	1,396	0.146	7	0.00 s	0.01 s	< 0.01 s	< 0.01 s	Communication	
Copnet FB ¹	800	12,836	0.003	4	0.02 s	0.01 s	< 0.01 s	< 0.01 s	Facebook	
FB Reed98 ²¹	962	37,624	0.006	3	0.10 s	0.03 s	0.01 s	0.01 s	Facebook	
Arenas email ¹⁹	1,133	10,902	0.023	5	0.02 s	0.02 s	< 0.01 s	< 0.01 s	Communication	
Network science ¹⁹	1,461	5,484	0.406	6	0.01 s	0.01 s	0.01 s	0.01 s	Co-autorship	
FB Simmons81 ²¹	1,518	65,976	0.006	3	0.19 s	0.04 s	0.01 s	0.01 s	Facebook	
DNC emails ¹⁹	1,893	8,770	0.640	4	0.02 s	0.37 s	< 0.01 s	< 0.01 s	Facebook	
Moreno health 19	2,539	20,910	0.002	5	0.03 s	0.07 s	< 0.01 s	< 0.01 s	Human social	
FB Wellesley22 ²¹	2,970	189,798	0.000	4	0.71 s	0.12 s	0.03 s	0.03 s	Facebook	Twin-unique:
Bitcoin alpha ²¹	3,783	28,248	0.241	5	0.07 s	0.56 s	< 0.01 s	< 0.01 s	Online social (trust)	•
GRQC collab. ²²	5,242	28,968	0.219	8	0.08 s	0.19 s	< 0.01 s	< 0.01 s	Co-autorship	unique or
FB Carnegie49 ²¹	6,637	499,934	0.004	4	2.47 s	0.81 s	0.09 s	0.09 s	Facebook	 all candidates are
Pajek Erdős ¹⁹	6,927	23,700	0.658	6	0.04 s	0.32 s	< 0.01 s	< 0.01 s	Co-autorship	
DT interaction ²³	7,341	30,276	0.448	12	0.09 s	5.44 s	0.01 s	0.01 s	Bio	twin
DG assoc. ²³	7,813	42,714	0.469	8	0.21 s	5.69 s	< 0.01 s	0.01 s	Bio	
FB GWU54 ²¹	12,193	939,056	0.003	4	4.62 s	1.63 s	0.14 s	0.14 s	Facebook	
Anybeat ²¹	12,645	98,264	0.447	5	0.64 s	1.3 h	0.01 s	0.01 s	Online social	
CE-CX ²¹	15,229	491,904	0.011	6	1.75 s	2.71 s	0.07 s	0.07 s	Bio	
Astro Physics ²¹	18,771	396,100	0.089	6	1.83 s	1.013 s	0.05 s	0.05 s	Co-autorship	
FB BU10 ²¹	19,700	1,275,056	0.003	4	5.45 s	2.58 s	0.19 s	0.19 s	Facebook	
FB Uillinois ²¹	30,664	2,097,148	0.001	4	10.04 s	4.64 s	0.31 s	0.31 s	Facebook	
Enron email ²²	36,692	367,662	0.382	6	1.51 s	2.32 m	0.05 s	0.05 s	Communication	
FB Penn ²¹	41,536	2,724,440	0.001	4	15.19 s	9.57 s	0.41 s	0.42 s	Facebook	
FB wall 2009 ¹⁹	46,952	366,824	0.083	8	0.69 s	2.76 s	0.05 s	0.07 s	Communication	
Brightkite ²¹	58,228	428,156	0.171	8	1.13 s	22.94 s	0.07 s	0.09 s	Online social	
The marker cafe ²⁴	69,413	3,289,686	0.168	5	1.12 m	19.83 m	0.58 s	0.60 s	Human contact	
Slashdot zoo ¹⁹	79,116	935,462	0.207	7	5.05 s	8.86 m	0.16 s	0.20 s	Online social	
Twitter ¹⁹	465,018	1,667,079	0.783	5	1.11 m	6.9 h	0.30 s	0.33 s	Online social	
DBLP ¹⁹	182,702	16,689,229	0.189	10	1.08 m	1.8 h	21.80 s	22.53 s	Co-autorship	
Flixster ¹⁹	2,523,387	15,837,601	0.584	8	3.54 m	9.6 h	15.59 s	16.05 s	Online social	
Youtube ¹⁹	3,223,590	18,750,747	0.335	15	16.26 m	>1 week	32.94 s	33.98 s	Online social	

Incorporating Twin-uniqueness

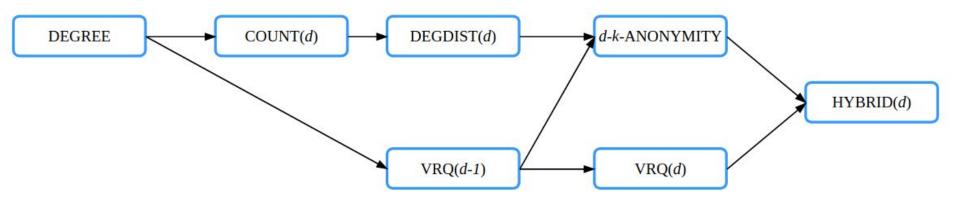


 $C_1 \rightarrow Large increase twin-uniqueness compared to uniqueness$

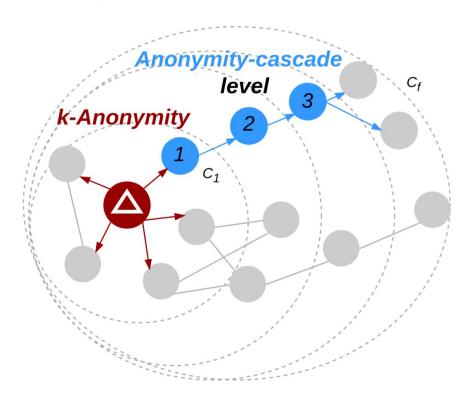
Measures

	d	= 1	d:	d = 2		
	Reach	Value	Reach	Value		
Degree [26, 61–63]	•	2	-	=		
Count [64]	•	(3, 3)		(5, 6)		
Degree distribution [64]	< 1	$\{2, 2, 2\}$		$\{2, 2, 2, 3, 3\}$		
<i>l-k</i> -Anonymity [10, 11, 65, 66]	•	< 1		•		
VRQ [27, 38]		$\{2, 3, 3\}$	•	$\{2, 2, 3, 3, 3\}$		
Hybrid [67]	~	{2,3,3}	•	{2,2,3,3,3}		

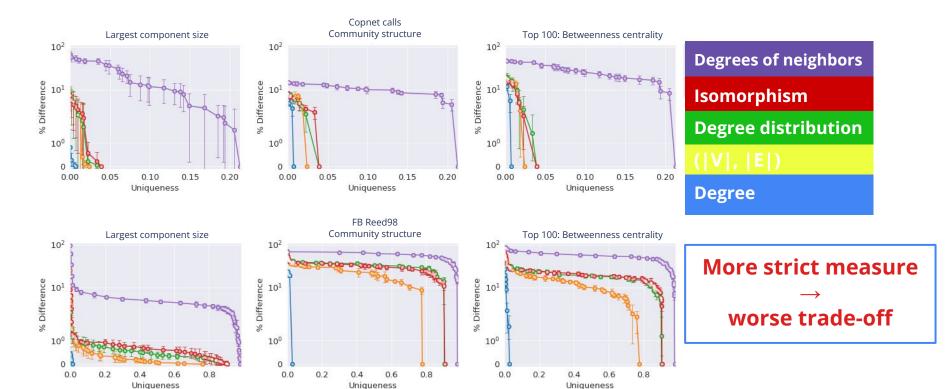
Measures: strictness



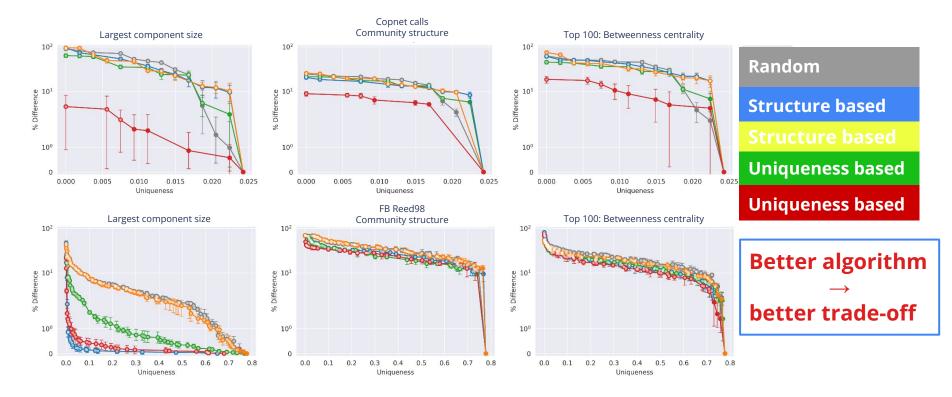
Anonymity-cascade



Anonymity vs. utility: measures



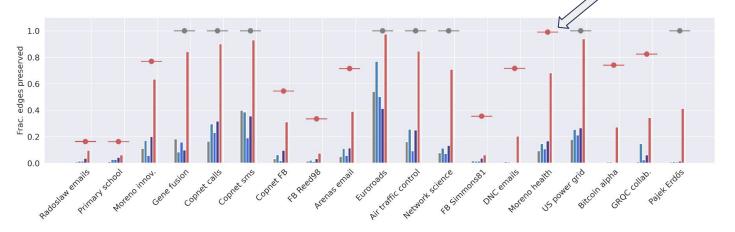
Anonymity vs. utility: algorithms



95% anonymous







Random

Structure based

Uniqueness based

|V|: ~100 - 6,927 |E|: ~300 - 11,850

For more details see our preprint!

