## war pact network model: generative model of networks that shrink

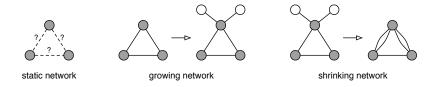
Lovro Šubelj University of Ljubljana Faculty of Computer and joint work with

Luka Naglić University of Zagreb Faculty of Science

#### **EUSN '19**

#### network models

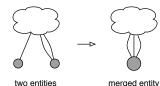
(soa) network models as baseline, explanation & generation
 (existing) majority for static or growing networks [ER59, Pri76]
 (missing) generative models of shrinking networks [KNB08]



[ER59] Erdős & Rényi (1959) On random graphs I. Publ. Math. Debrecen 6, 290-297.
 [Pri76] Price (1976) A general theory of bibliometric and other cumulative. . . J. Am. Soc. Inf. Sci. 27(5), 292-306.
 [KNB08] Kejžar et al. (2008) Probabilistic inductive classes of graphs. J. Math. Sociol. 32(2), 85-109.

### shrinking models

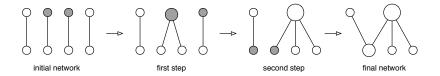
(intuition) entities/nodes often merge in real world/network (which) merged nodes/entities are random, hubs, isolates etc.



(wars) nations/alliances form pact or one occupies other •
 (trade) countries form alliance or companies after merger
 (Bitcoin) cryptocurrency addresses owned by same user
 (Internet) autonomous systems merge their traffic

war pact model

(model) shrinking network with n nodes & m edges



(initialize) create perfect matching on 2m nodes
 (select) select nodes at random, preferentially etc.
 (shrink) merge nodes by rewiring their edges
 (loop) continue until network has n nodes

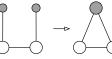
#### model **details**

#### (shrink) merging nodes at distance d creates d-cycle









edge with d = 1 self-edge

-edge p

path of length d = 2 parallel edges

path of length d = 3

triangle

(model) war pact is parameter-free except *n* nodes & *m* edges (initialize) create perfect matching, random graph or tree  $\circ$ (select) select nodes at random, by degree or degree<sup>-1</sup> •

### model pseudocode

```
input nodes n & edges m output graph G
```

- 1:  $H \leftarrow empty map$
- 2:  $G \leftarrow empty graph$
- 3: for  $i \in [1, m]$  do
- $4: \qquad H(i) \leftarrow i \& H(m+i) \leftarrow m+i$
- 5: add nodes H(i) & H(m+i) to G
- 6: add edge  $\{H(i), H(m+i)\}$  to G
- 7: while G has > n nodes do
- 8:  $h \leftarrow random(H)$
- 9:  $i \leftarrow random([1, 2m])$
- 10: if  $h \neq H(i)$  & edge  $\{h, H(i)\} \notin G$  then
- 11: merge nodes h & H(i) in G12:  $H(i) \leftarrow h$

13: return G

map of nodes' hashesempty war pact graph

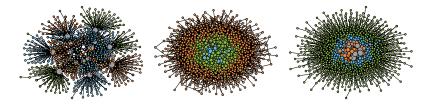
map nodes to hashes
 add nodes to graph
 add edges to graph

select random nodeselect node by degree

merge selected nodesunify nodes' hashes

#### model networks

(layout) node selection impacts (modular) structure [Pei18]

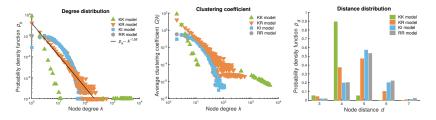


(left) both nodes are selected by degree
(middle) nodes selected by degree & degree<sup>-1</sup>
(right) nodes selected by degree & at random

[Pei18] Peixoto (2018) Bayesian stochastic blockmodeling. e-print arXiv:1705.10225v7, 1-44.

### model selection

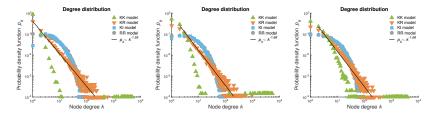
#### (structure) node selection impacts scale-free/small-world



(KK model) both are nodes selected by degree
 (KR model) nodes selected by degree & at random
 (KI model) nodes selected by degree & degree<sup>-1</sup>
 (RR model) both nodes are selected at random

### model initialization

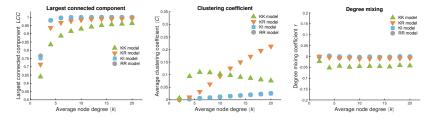
#### (structure) model initialization has no apparent impact



(left) networks initialized by perfect matching(middle) networks initialized by random graph(right) networks initialized by random tree

### model evolution

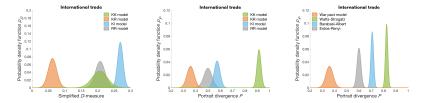
#### (structure) model evolution when increasing node degree $\langle k \rangle$



(left) emergence of giant component LCC when increasing (k)
 (middle) increasing node clustering (C) when increasing (k)
 (right) "fixed" degree mixing r when changing (k)

### model comparison

(network) international trade (i.e. food import & export) (models) war pact  $\gg$  small-world, scale-free & random graphs

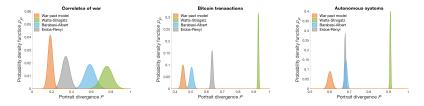


#### (left) simplified D-measure [SCDPMR17] (right) portrait divergence P [BB19]

[SCDPMR17] Schieber et al. (2017) Quantification of network structural dissimilarities. Nat. Commun. 8, 13928.
 [BB19] Bagrow & Bollt (2019) An information-theoretic, all-scales approach to comparing. . . Appl. Netw. Sci. 4, 45.

### model validation

#### (networks) national wars, Bitcoin transactions & Internet map (models) war pact ≫ small-world, scale-free & random graphs



#### (measure) portrait divergence P [BB19]

[BB19] Bagrow & Bollt (2019) An information-theoretic, all-scales approach to comparing. . . Appl. Netw. Sci. 4, 45.

#### model structure

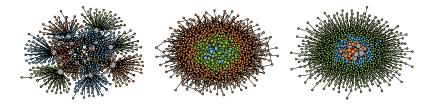
(size) model reproduces nodes n & edges m by design (connectivity) model well reproduces giant component *LCC* (distance) model well reproduces distance  $\langle d \rangle \&$  diameter  $d_{max}$ 

	n	т	$\langle k \rangle$	LCC	$\langle C \rangle$	$\langle d \rangle$	d <sub>max</sub>
Correlates of war	41	54	2.63	87.8%	0.28	2.58	8
	41	54	2.63	90.2%	0.06	2.64	7
International trade	130	3 7 3 0	57.38	100.0%	0.50	2.24	5
	130	3 7 3 0	57.38	100.0%	0.53	2.17	5
Bitcoin transactions	1 288	6 2 3 6	9.68	98.8%	0.33	2.83	9
	1 288	6 236	9.68	98.0%	0.13	3.08	7
Autonomous systems	3213	11 248	7.00	100.0%	0.18	3.77	9
	3 2 1 3	11 248	7.00	98.3%	0.03	3.62	9

(clustering) model often underestimates node clustering  $\langle C \rangle$ 

#### model conclusions

(novel) simple model of networks that shrink
 (others) in contrast to classic static & growing models
 (networks) model well reproduces structure except clustering



(question) growing or shrinking models more "reasonable"?(future) combined model, other networks & analytical results

# thank you!

# arXiv:1909.00745v1

Naglić & Šubelj (2019) War pact model of shrinking networks. PLoS ONE, under review.

Lovro Šubelj University of Ljubljana

lovro.subelj@fri.uni-lj.si
http://lovro.lpt.fri.uni-lj.si

joint work with

Luka Naglić University of Zagreb

lu.naglic@gmail.com
http://www.pmf.unizg.hr