A stochastic agent-based model of pathogen propagation in dynamic multi-relational social networks

Bilal Khan, Kirk Dombrowski, and Mohamed Saad, Journal of Transactions of Society Modeling and Simulation International, SAGE, 2014).

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Overview

- Network Simulation
- ERGM vs ABM
- Model
- Design of MABUSE, the system
- Validation

Network Simulation

- What is a network?
- What is a dynamic network?
- Networks in the real world
- Networks in simulation

Main network simulation strategies

- ERGM (Exponential Random Graph Modeling) Hunter DR, Handcock MS, Butts CT, Goodreau SM, Morris M. ergm: A Package to Fit, Simulate and Diagnose Exponential-Family Models for Networks. Journal of statistical software. 2008;24(3):nihpa54860.
- ABM (this class)

Exponential Random Graph Modeling (ERGM)



Exponential Random Graph Modeling (ERGM)



Agent Based Model Simulation (ABM)



Agent Based Model Simulation (ABM)



ERGM vs ABM

- ERGM gives you realistic network-level guarantees but lacks realistic individual agency
 - The network gets to change realistically over time
 - What each agent is doing could be unrealistic
- ABM gives you lack realistic individual agency but lacks realistic network-level guarantees
 - Each agent gets to change realistically over time
 - What the network is doing could be unrealistic

MABUSE

- MABUSE gives you
 - The network gets to change realistically over time

• Each agent gets to change realistically over time

Outline of MABUSE Model

INITIALIZATION

- 1. Nodes are made (including some infected)
 - 2. Edges are made

SIMULATION RUN

- 3. Edges change over time
- 4. Risk acts happen with neighbors
- 5. Nodes come and go

Nodes













Nodes with multiple





















Population









Population



Population



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How individuals choose who to form relationships with?



Bivariate Distribution

Gender Bivariate

	Male	Female
Male	0.8	0.2
Female	0.25	0.75

Age Bivariate

	Young	Old
Young	0.6	0.4
Old	0.3	0.7

Aggregator

- Gender: 0.8
- Age: 0.3

(from ERGM)



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(from ERGM)

Calculate Propensity



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- Age: 0.3

(from ERGM)

Propensity 1-2 = 0.25*0.8 + 0.4*0.3 = 0.32

Calculate Propensity



Calculate Propensity





Random Selection



5 0.14 0.17 0.33 99 0.36 10

Random Selection



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Dynamic edges of different types

Every node has an "ideal degree"

Univariate Distribution



Every Relationship has a "longevity"





Day 7: Relationship 1-2 ends



Day 7: Relationship 1-2 ends







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Every node has a "risk interval"

Univariate Distribution



What are the people doing? Risky things











The "natural history" of disease





=0.004 in acute =0.00004 in chronic

The "natural history" of disease





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Univariate Distribution





Day 26: Person 1 dies



Day 26: Person 1 dies



Day 26: Person 1 dies



Now all these people have Actual degree < Ideal degree...

... and the population shrank.



New Relationships are made



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Now, putting it all together...

One layer, one pathogen scenario

Node Longev Minimum (days) Maximum (days)	50 100		Gender Male Female	0.7		Age Young Old	0.4
Risk Behavior (F Low 100 Medium 100 High 100	Per Week)	Netwo Ideal Degree Low 0 Medium 1 High 0	rk Layer	Dru Edge Longe Short Max Short Min Long Max Long Min	g layer evity (Years) 30 1 1825 30	¢	Transmission probability #1

Multi-layer scenario

Node Longevity	Gender		Age	
Minimum 50 (days)	Male	0.7	Young	0.4
Maximum 100 (days)	Female	0.3	Old	0.6
Risk Behavior (Per Week) Low 100 Medium 100 High 100	Ideal Degree (Partners) Low 0 Medium 1 High 0	H1 Drug layer Edge Longevity (Years) Short Max 30 Short Min 1 Long Max 1825 Long Min 30	*	Transmission probability #1
Risk Behavior (Per Week)Low100Medium100High100	Ideal Degree (Partners) Low 0 Medium 1 High 0	#2Sex layerEdge Longevity (Years)Short Max30Short MinLong Max1825Long Min30	¢	Transmission probability #1

Multi-pathogen scenario

Node Longevity Minimum (days) Maximum (days)	5 0	Gender Male Female	0.7	Age Young Old	0.4
Risk Behavior (Per Low 100 Medium 100 High 100	r Week) Ide	twork Layer eal Degree (Partners) ow 0 ledium 1 igh 0	Drug layerEdge Longevity (Years)Short Max30Short Min1Long Max1825Long Min30	♪ ◆	Transmission probability #1 Transmission probability #1

Multi-pathogen Multi-layer scenario

Node Longevity		Gender		Age	
Minimum 50 (days)		Male	0.7	Young	0.4
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Risk Behavior (Per Wee Low 100 Medium 100 High 100	ek) Ideal Degree	(Partners)	JDrug layerEdge Longevity (Years)Short Max30Short Min1Long Max1825Long Min30	、 、 、 、 、 、 、 、 、 、 、 、 、	Transmission probability #1 Transmission probability #1
	Networ	rk Layer #	2 Sex layer	•	

#1

#1

Risk Behavior (I	Per Week)	Ideal De	egree (Partners)	Edge Longe	vity (Years)		Transmission probability #1
Low 100		Low	0	Short Max	30		
Medium 100		Medium	1	Short Min	1	adat	T
High 100		High	0	Long Max	1825	H	probability #1
				Long Min	30	· · · · · · · · · · · · · · · · · · ·	

Design of MABUSE

Key features

- Dynamic network simulation
 - Multiple (dynamic) node attributes
 - Multiple (dynamic) edge layers
 - Multiple simultaneous pathogens



- Stochastically defined node agency
 - E.g. risk activity along network edges on different layers is what drives the flow of multiple pathogen types

How is the implementation going to be validated?

- Validate Network Dynamism
- Validate Pathogen Flow

Validation of Network Simulation

- 1. Take a real network. Compute statistics
 - A. Structure (univariates and bivariates)
 - B. Dynamism (ideal degrees, longevity)
- 2. Use 1A+1B to do a 10 year simulation. At the end, Compute statistics
 - A. Structure (univariates and bivariates)
 - B. Dynamism (ideal degrees, longevity)
- Is 1A "close" to 2A? Is 1B "close" to 2B?

Use Shannon-Jensen divergence to measure the "distance" between distributions










Results



Time

Validation of Pathogen

- Take Base Scenario
 - Single Layer
 - Single Pathogen
- Compare Base Scenario with the following Artificial Scenarios

Number of Layers	Pathogen I prevalence at 60 months (%)
l = 2 (and p = 1)	70
l=3 (and $p=1$) l=4 (and $p=1$)	89
Base scenario ($p = I$ and $I = I$)	42

 Table 5.
 Scenario I: Multiple layers, one pathogen.



Table 6. Scenario 2: Multiple pathogens, one layer.

Number of pathogens	Average prevalence at 60 months (across all þ pathogens) (%)	Ave. pairwise correlation of pathogen occurrence
p = 2 (and $l = 1$) p = 3 (and $l = 1$) p = 4 (and $l = 1$) Base scenario ($p = 1$ and $l = 1$)	42 43 44 42	0.86 0.76 0.72

Table 7.	Scenario	3:	Multiple	layers,	multiple	non-interacting
pathogens						

Number of pathogens p Number of layers l	Average prevalence at 60 months (across all <i>p</i> pathogens) (%)	Ave. pairwise correlation of pathogen occurrence
p = l = 2 p = l = 3 p = l = 4 Base scenario (p = l = 1)	42 43 42 42	0.12 0.11 0.08

 Table 8.
 Scenario 4: Multiple layers, multiple interacting pathogens.

Number of Pathogens p Number of Layers I	Average prevalence at 60 months (across all p pathogens)(%)
p = l = 2	22
p = l = 3	15
p = l = 4	12
Base Scenario ($p = l = 1$)	42



Multi Actor-Based Universal Simulation Engine

• MABUSE is stochastic, agent-based, discrete event simulator for epidemics in dynamic networks.



 The simulator runs on specialized highperformance hardware and is capable of running simulations of dynamic networks having 1,000,000+ nodes, and handling approximately 800 million discrete events per second.

Conclusion

- MABUSE allows us to simulate
 - Multiple (dynamic) node attributes
 - Multiple (dynamic) edge layers
 - Multiple simultaneous pathogens
- A realistic dynamic network
 - From the POV of an agent (micro, ABM-style)
 - From POV of the network (macro, ERGM-style)
- Model where each agent takes into account its own properties and the properties of other agents to make connections, and act on those connections

Thank you

Any Questions?

Jensen-Shannon Divergence



$$SJ(D1, D2) = \frac{1}{2} [KL(D1, M) + KL(D2, M)]$$

$$\mathsf{KL}(D1, M) = \sum_{i=M,F} D1(i) * \log(\frac{D1(i)}{Mid(i)}) = 0.55 * \log\left(\frac{0.55}{0.65}\right) * 0.45 * \log(\frac{0.44}{0.35})$$