

Chaotic social contagion

from zombies to hipsters

Kameron Decker Harris
with Peter Dodds and Chris Danforth

UVM Math

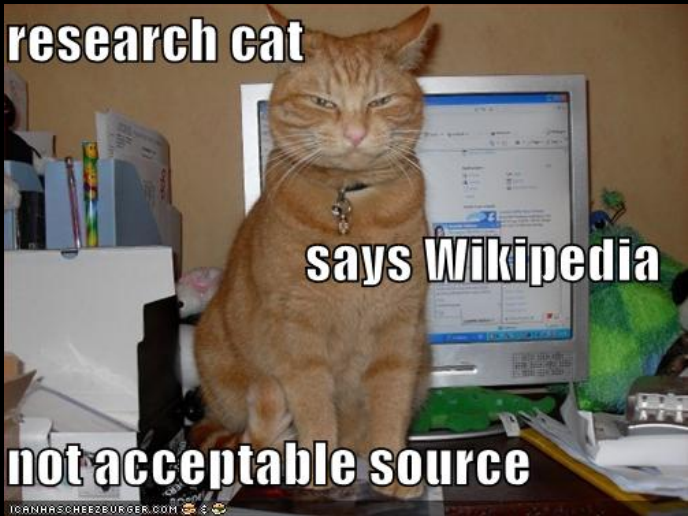
April 19, 2012



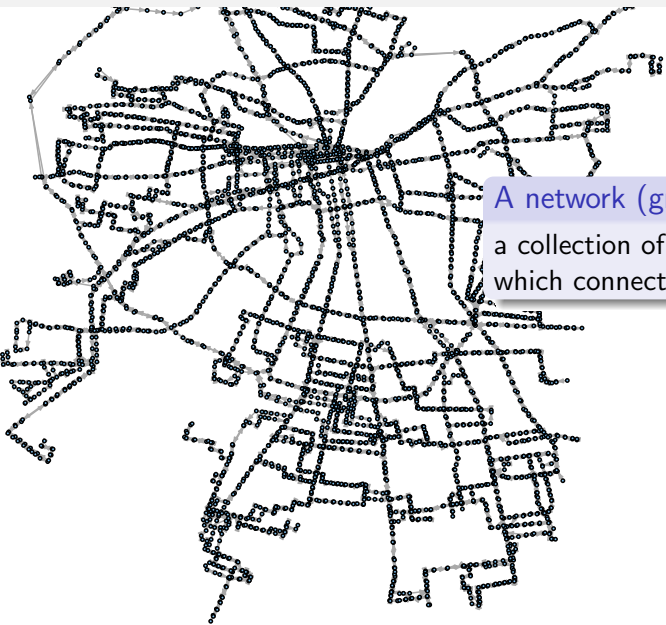
You may think



I think



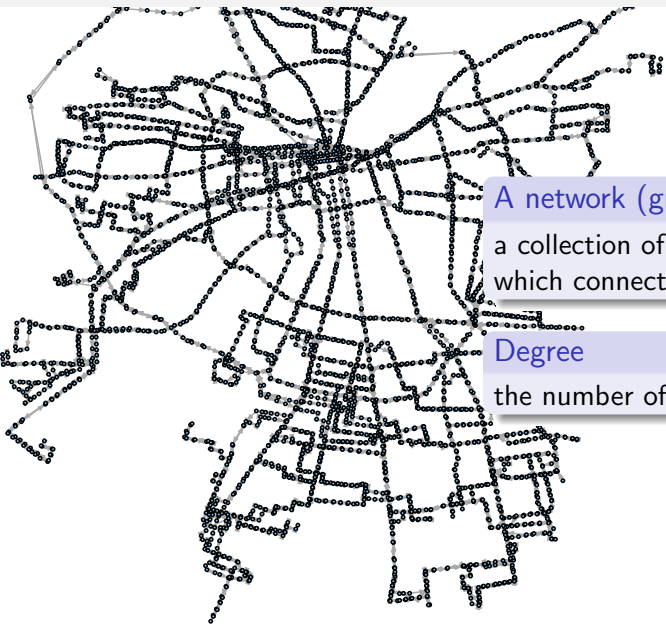
Networks: quick definitions



A network (graph) is

a collection of **nodes** and **links**,
which connect the nodes.

Networks: quick definitions



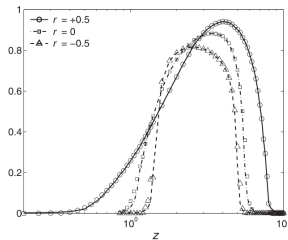
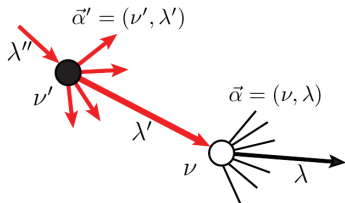
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Degree

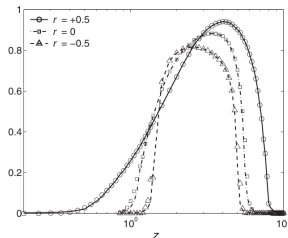
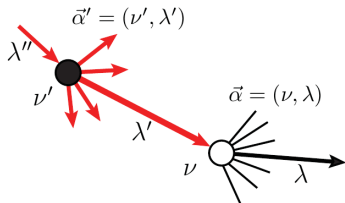
the number of links incident a node

Contagion models



Typical features:

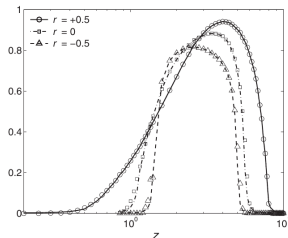
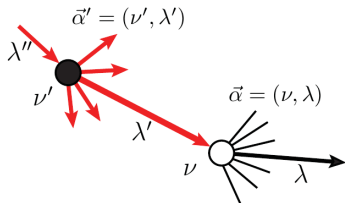
Contagion models



Typical features:

- **Network structure** or fully mixed

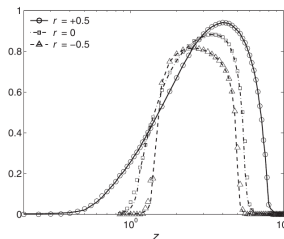
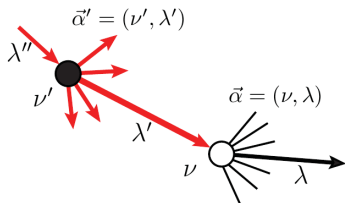
Contagion models



Typical features:

- **Network structure** or fully mixed
- Nodes turn on according to **response function** f and stay on

Contagion models



Typical features:

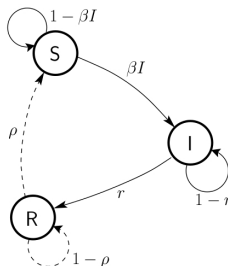
- **Network structure** or fully mixed
- Nodes turn on according to **response function** f and stay on
- **Percolation transition**. Varies with network parameters, contagiosity (f), or initial seed size

Percolation models & Networks (cont'd)

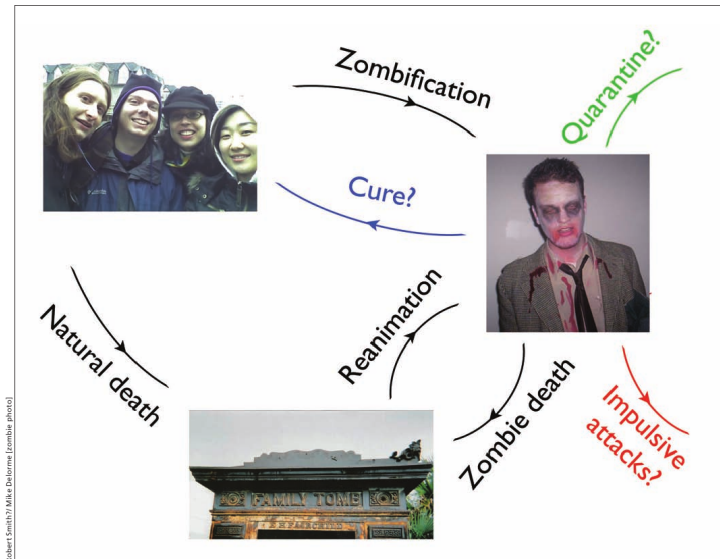
SIR

Most models are

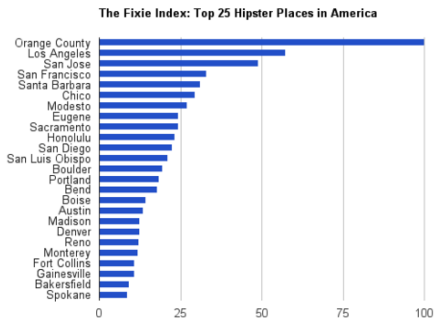
- Susceptible
- Infected
- Removed



Or SZR



Other possible extensions



Fixie Index

source: <http://blog.priceonomics.com/post/16013457968/the-fixie-bike-index>

- Social contagion → trends?
- Volatility: not something we get from SIR

Chaotic contagion

Main idea

Nodes turn on and off (contact process, SIS, SZS, etc.)

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How we do this

Non-monotonic response functions

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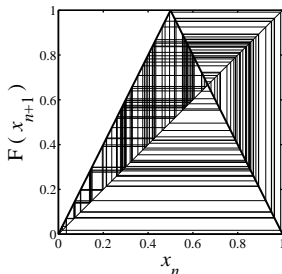
How we do this

Non-monotonic response functions

Ingredients:

Nodes like to imitate up to a limit — don't want to be like everyone else.

An example f :



Chaotic contagion

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How we do this

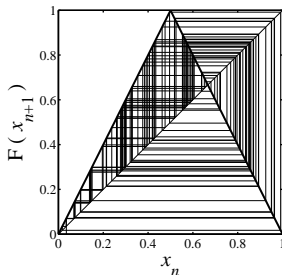
Non-monotonic response functions

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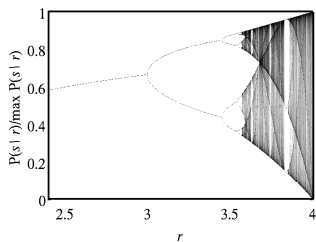
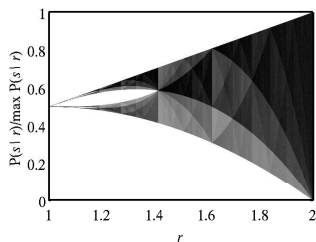
Nodes like to imitate up to a limit — don't want to be like everyone else.

(Hipsters!!!)

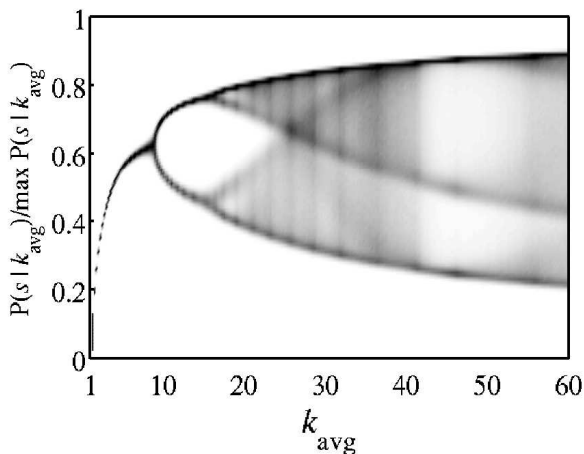
An example f :



Crazy behavior



Stochastic response functions:



How to understand this?

Mean field theory

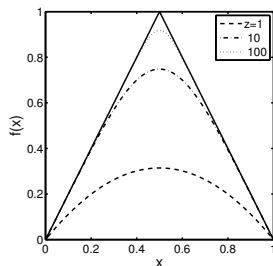
This leads to a 1d, nonlinear map:

$$\rho_{t+1} = \alpha g(\rho_t | P, f) + (1 - \alpha)\rho_t \quad (1)$$

Rigorous analytic results

Result 1

As $\langle k \rangle_P = z$ grows, for reasonable P , $g \nearrow f$
(More connected \implies dynamics close to f)



Rigorous analytic results

Result 1

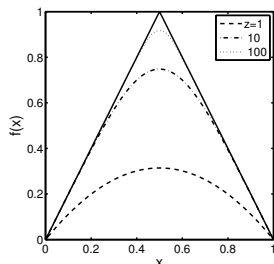
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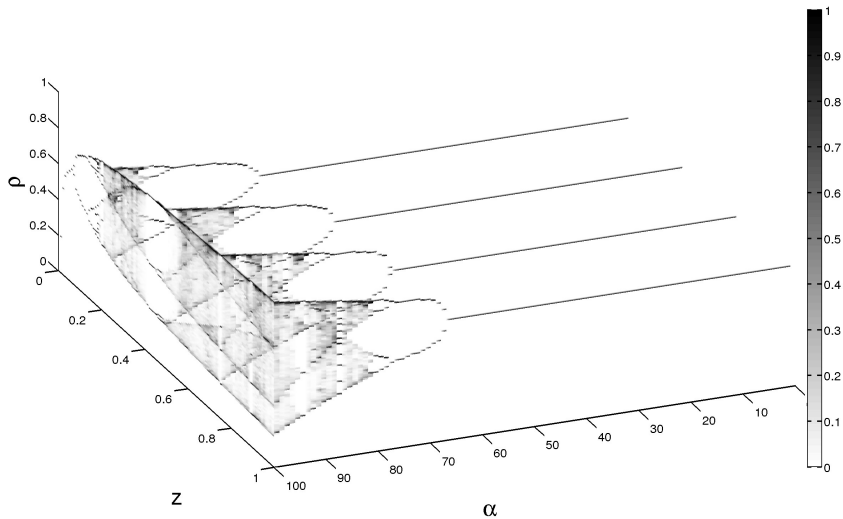
Result 2

For

- quenched Poisson random networks (not mean field)
- in the dense limit

the dynamics also approach f





Conclusions, to-dos

- Look at individual node dynamics
- Quenched networks, quenched response functions

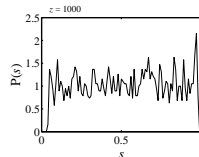
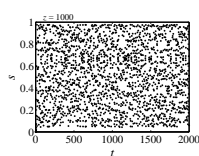
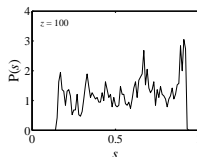
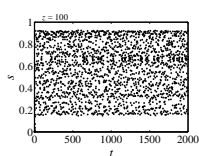
Main conclusion

Academics can study anything
(including zombies, hipsters, swearing on Twitter...)

Acknowledgments

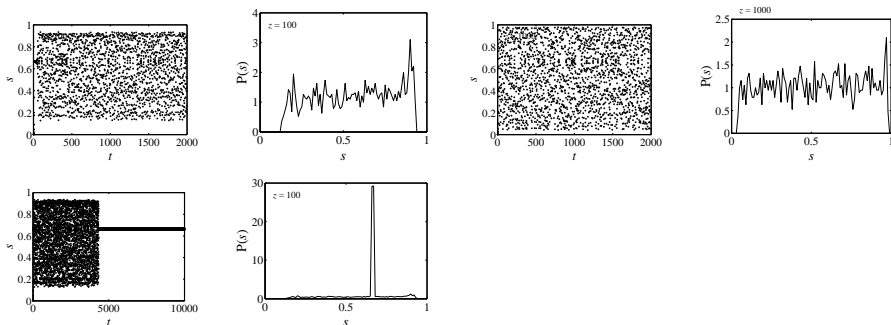
- advisors Peter Dodds and Chris Danforth
- Josh Payne
- the whole Computational Story Lab/Onehappybird team
- Andi Elledge (you rock!)
- the VACC, UVM Complex Systems Center, UVM Math & Stats
- NASA, NSF
- ...and those of you who came to see this

Invariant densities—stochastic response functions



Trying out higher values of $\langle k \rangle \dots$

Invariant densities—deterministic response functions



Trying out higher values of $\langle k \rangle$...