

Generalized Contagion

Principles of Complex Systems

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Generalized Model of Contagion

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Basic questions about contagion

- ▶ How many types of contagion are there?
- ▶ How can we categorize real-world contagions?
- ▶ Can we connect models of disease-like and social contagion?

Some (of many) issues

- ▶ Disease models assume independence of infectious events.
- ▶ Threshold models only involve proportions:
 $3/10 \equiv 30/100$.
- ▶ Threshold models ignore exact sequence of influences
- ▶ Threshold models assume immediate polling.
- ▶ Mean-field models neglect network structure
- ▶ Network effects only part of story:
media, advertising, direct marketing.

Generalized model—ingredients

- ▶ Incorporate memory of a contagious element ^[1, 2]
- ▶ Population of N individuals, each in state S, I, or R.
- ▶ Each individual randomly contacts another at each time step.
- ▶ ϕ_t = fraction infected at time t
= probability of contact with infected individual
- ▶ With probability p , contact with infective leads to an exposure.
- ▶ If exposed, individual receives a dose of size d drawn from distribution f . Otherwise $d = 0$.

$S \Rightarrow I$

- ▶ Individuals ‘remember’ last T contacts:

$$D_{t,i} = \sum_{t'=t-T+1}^t d_i(t')$$

- ▶ Infection occurs if individual i 's ‘threshold’ is exceeded:

$$D_{t,i} \geq d_i^*$$

- ▶ Threshold d_i^* drawn from arbitrary distribution g at $t = 0$.

Generalized model—ingredients

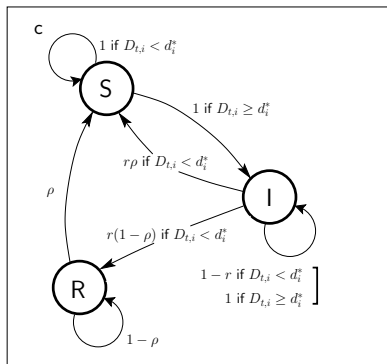
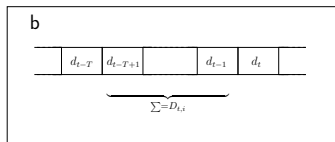
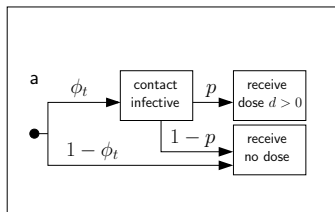
$$I \Rightarrow R$$

When $D_{t,i} < d_i^*$,
individual i recovers to state R with probability r .

$$R \Rightarrow S$$

Once in state R, individuals become susceptible again
with probability ρ .

A visual explanation



Important quantities:

$$P_k = \int_0^{\infty} dd^* g(d^*) P \left(\sum_{j=1}^k d_j \geq d^* \right) \text{ where } 1 \leq k \leq T.$$

P_k = Probability that the threshold of a randomly selected individual will be exceeded by k doses.

e.g.,

P_1 = Probability that one dose will exceed the threshold of a random individual
= Fraction of most vulnerable individuals.

Generalized model—heterogeneity, $r = 1$

Fixed point equation:

$$\phi^* = \sum_{k=1}^T \binom{T}{k} (p\phi^*)^k (1 - p\phi^*)^{T-k} \underline{P}_k$$

Expand around $\phi^* = 0$ to find Spread from single seed if

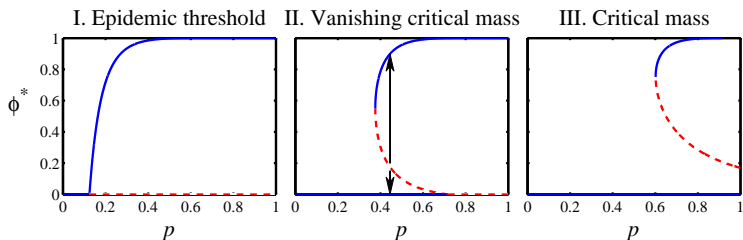
$$pP_1 T \geq 1$$

$$\Rightarrow p_c = 1/(TP_1)$$

Example configuration:

- ▶ Dose sizes are lognormally distributed with mean 1 and variance 0.433.
- ▶ Memory span: $T = 10$.
- ▶ Thresholds are uniformly set at
 1. $d_* = 0.5$
 2. $d_* = 1.6$
 3. $d_* = 3$
- ▶ Spread of dose sizes matters, details are not important.

Heterogeneous case—Three universal classes



- ▶ Epidemic threshold: $P_1 > P_2/2, p_c = 1/(TP_1) < 1$
- ▶ Vanishing critical mass: $P_1 < P_2/2, p_c = 1/(TP_1) < 1$
- ▶ Pure critical mass: $P_1 < P_2/2, p_c = 1/(TP_1) > 1$

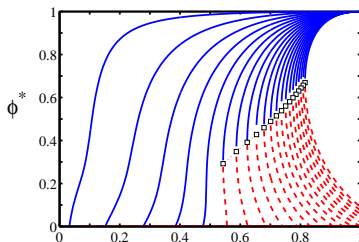
Calculations—Fixed points for $r < 1$, $d^* = 2$, and $T = 3$

$$\text{F.P. Eq: } \phi^* = \Gamma(p, \phi^*; r) + \sum_{i=d^*}^T \binom{T}{i} (p\phi^*)^i (1 - p\phi^*)^{T-i}.$$

$$\Gamma(p, \phi^*; r) = (1 - r)(p\phi)^2(1 - p\phi)^2 + \sum_{m=1}^{\infty} (1 - r)^m (p\phi)^2 (1 - p\phi)^2 \times \\ \left[\chi_{m-1} + \chi_{m-2} + 2p\phi(1 - p\phi)\chi_{m-3} + p\phi(1 - p\phi)^2\chi_{m-4} \right]$$

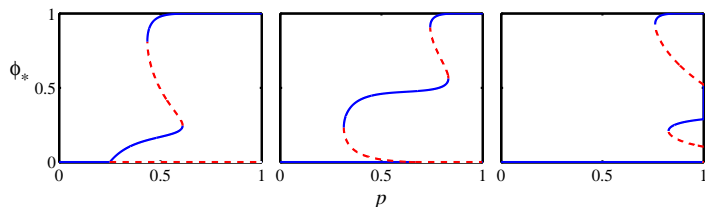
$$\text{where } \chi_m(p, \phi^*) = \sum_{k=0}^{\lfloor m/3 \rfloor} \binom{m-2k}{k} (1 - p\phi^*)^{m-k} (p\phi^*)^k.$$

Now allow $r < 1$:



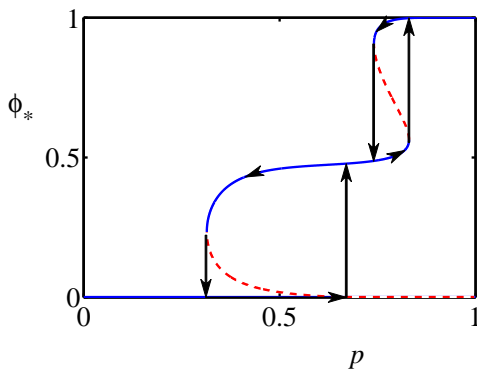
II-III transition generalizes: $p_c = 1/[P_1(T + \tau)]$
(I-II transition less pleasant analytically)

More complicated models



- Due to heterogeneity in individual thresholds.
- Same model classification holds: I, II, and III.

Hysteresis in vanishing critical mass models



Generalized model—heterogeneity, $r \leq 1$

II-III transition generalizes:

$$p_c = 1/[P_1(T + \tau)]$$

where $\tau = 1/r =$ expected recovery time

- ▶ Memory is crucial ingredient.
- ▶ Three universal classes of contagion processes:
 - I. Epidemic Threshold
 - II. Vanishing Critical Mass
 - III. Critical Mass
- ▶ Dramatic changes in behavior possible.
- ▶ To change kind of model: 'adjust' memory, recovery, fraction of vulnerable individuals (T , r , ρ , P_1 , and/or P_2).
- ▶ To change behavior given model: 'adjust' probability of exposure (p) and/or initial number infected (ϕ_0).

- ▶ If $pP_1(T + \tau) \geq 1$, contagion can spread from single seed.
- ▶ Key quantity: $p_c = 1/[P_1(T + \tau)]$
- ▶ Depends only on:
 1. System Memory ($T + \tau$).
 2. Fraction of highly vulnerable individuals (P_1).
- ▶ **Details unimportant** (Universality):
Many threshold and dose distributions give same P_k .
- ▶ Most vulnerable/gullible population may be more important than small group of super-spreaders or influentials.

Future work/questions

- ▶ Do any real diseases work like this?
- ▶ Examine model's behavior on networks
- ▶ Media/advertising + social networks model
- ▶ Classify real-world contagions



P. S. Dodds and D. J. Watts.

Universal behavior in a generalized model of contagion.

Phys. Rev. Lett., 92:218701, 2004. [pdf](#) (田)



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