

# Contagion

## Complex Networks

### CSYS/MATH 303, Spring, 2011

Prof. Peter Dodds

Department of Mathematics & Statistics  
Center for Complex Systems  
Vermont Advanced Computing Center  
University of Vermont

Contagion

Basic Contagion  
Models

Global spreading  
condition

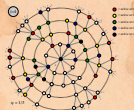
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Outline

## Basic Contagion Models

## Global spreading condition

## Social Contagion Models

Network version

All-to-all networks

Theory

## References

Contagion

Basic Contagion  
Models

Global spreading  
condition

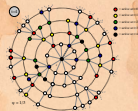
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given spreading mechanism on a given network, what's the probability that there will be global spreading?
  2. If spreading does take off, how far will it go?
  3. How do the details of the network affect the outcome?
  4. How do the details of the spreading mechanism affect the outcome?
  5. What if the seed is one or many nodes?
- ▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

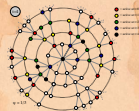
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
  2. If spreading does take off, how far will it go?
  3. How do the details of the network affect the outcome?
  4. How do the details of the spreading mechanism affect the outcome?
  5. What if the seed is one or many nodes?
- ▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

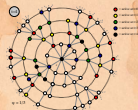
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
  2. If spreading does take off, how far will it go?
  3. How do the details of the network affect the outcome?
  4. How do the details of the spreading mechanism affect the outcome?
  5. What if the seed is one or many nodes?
- ▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

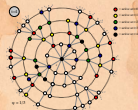
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
  2. If spreading does take off, how far will it go?
  3. How do the **details** of the **network** affect the outcome?
  4. How do the details of the spreading mechanism affect the outcome?
  5. What if the seed is one or many nodes?
- ▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

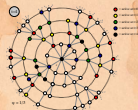
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
2. If spreading does take off, how far will it go?
3. How do the **details** of the **network** affect the outcome?
4. How do the **details** of the **spreading mechanism** affect the outcome?
5. What if the seed is one or many nodes?

▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

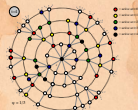
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
2. If spreading does take off, how far will it go?
3. How do the **details** of the **network** affect the outcome?
4. How do the **details** of the **spreading mechanism** affect the outcome?
5. What if the **seed** is one or many nodes?

▶ Next up: We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

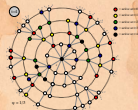
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References





# Contagion models

Some large questions concerning network contagion:

1. For a given **spreading mechanism** on a given network, what's the **probability** that there will be **global spreading**?
2. If spreading does take off, how far will it go?
3. How do the **details** of the **network** affect the outcome?
4. How do the **details** of the **spreading mechanism** affect the outcome?
5. What if the **seed** is one or many nodes?
  - ▶ **Next up:** We'll look at some fundamental kinds of spreading on generalized random networks.

Contagion

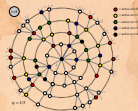
Basic Contagion Models

Global spreading condition

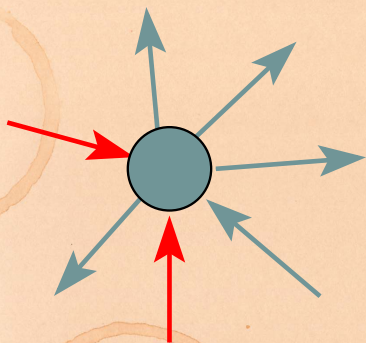
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Spreading mechanisms



■ uninfected  
■ infected

- ▶ **General spreading mechanism:**  
State of node  $i$  depends on history of  $i$  and  $i$ 's neighbors' states.
- ▶ Doses of entity may be stochastic and history-dependent.
- ▶ May have multiple, interacting entities spreading at once.

Contagion

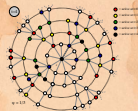
Basic Contagion Models

Global spreading condition

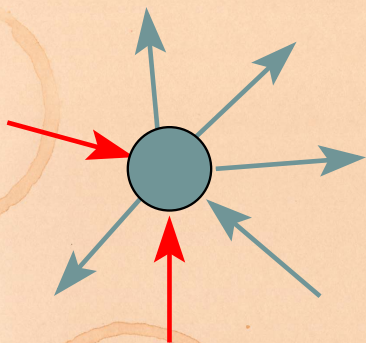
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Spreading mechanisms



■ uninfected  
■ infected

- ▶ **General spreading mechanism:**  
State of node  $i$  depends on history of  $i$  and  $i$ 's neighbors' states.
- ▶ **Doses** of entity may be stochastic and history-dependent.
- ▶ May have multiple, interacting entities spreading at once.

Contagion

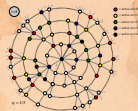
Basic Contagion Models

Global spreading condition

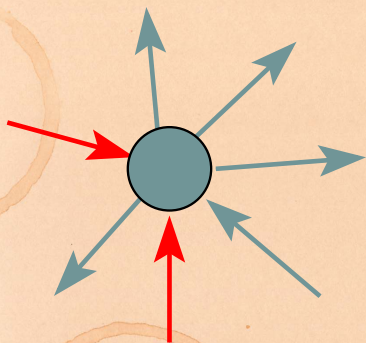
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Spreading mechanisms



■ uninfected  
■ infected

- ▶ **General spreading mechanism:**  
State of node  $i$  depends on history of  $i$  and  $i$ 's neighbors' states.
- ▶ **Doses** of entity may be stochastic and history-dependent.
- ▶ May have **multiple, interacting entities** spreading at once.

Contagion

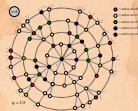
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Spreading on Random Networks

- ▶ For random networks, we know local structure is pure branching.
- ▶ Successful spreading is  $\therefore$  contingent on single edges infecting nodes.

- ▶ Focus on binary case with edges and nodes either infected or not.
- ▶ First big question: for a given network and contagion process, can global spreading from a single seed occur?

Contagion

Basic Contagion Models

Global spreading condition

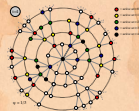
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Spreading on Random Networks

- ▶ For random networks, we know local structure is pure branching.
- ▶ Successful spreading is  $\therefore$  contingent on **single edges** infecting nodes.

- ▶ Focus on binary case with edges and nodes either infected or not.
- ▶ First big question: for a given network and contagion process, can global spreading from a single seed occur?

Contagion

Basic Contagion Models

Global spreading condition

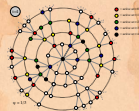
Social Contagion Models

Network version

All-to-all networks

Theory

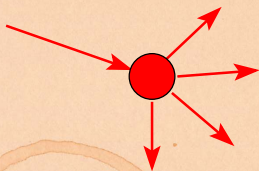
References



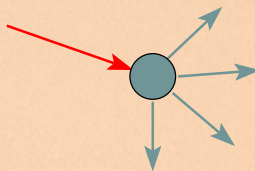
# Spreading on Random Networks

- ▶ For random networks, we know local structure is pure branching.
- ▶ Successful spreading is  $\therefore$  contingent on **single edges** infecting nodes.

Success



Failure:



- ▶ Focus on binary case with edges and nodes either infected or not.
- ▶ First big question: for a given network and contagion process, can global spreading from a single seed occur?

Contagion

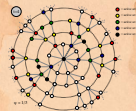
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

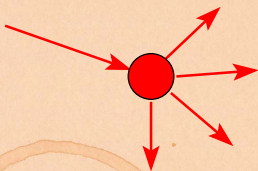
References



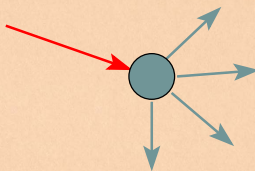
# Spreading on Random Networks

- ▶ For random networks, we know local structure is pure branching.
- ▶ Successful spreading is  $\therefore$  contingent on **single edges** infecting nodes.

Success



Failure:



- ▶ Focus on **binary** case with edges and nodes either infected or not.
- ▶ First big question: for a given network and contagion process, can global spreading from a single seed occur?

Contagion

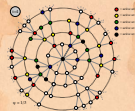
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References

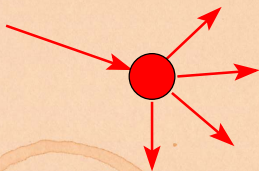




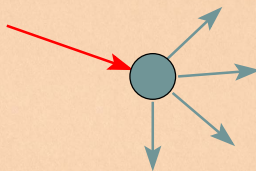
# Spreading on Random Networks

- ▶ For random networks, we know local structure is pure branching.
- ▶ Successful spreading is  $\therefore$  contingent on **single edges** infecting nodes.

Success



Failure:



- ▶ Focus on **binary** case with edges and nodes either infected or not.
- ▶ **First big question:** for a given network and contagion process, can global spreading from a single seed occur?

Contagion

Basic Contagion Models

Global spreading condition

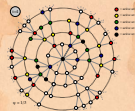
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

$$R = \sum_{k=0}^{\infty}$$

$$\frac{kP_k}{\langle k \rangle}$$

prob. of connecting to a degree  $k$  node

$$\cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}}$$

$$\cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle}$$

$$\cdot \underbrace{0}_{\text{\# outgoing infected edges}}$$

$$\cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

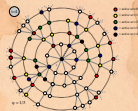
Basic Contagion Models

Global spreading co

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{\# outgoing infected edges}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

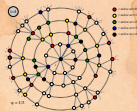
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{Prob. of no infection}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

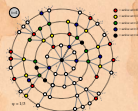
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{\# outgoing infected edges}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

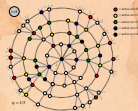
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{Prob. of no infection}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

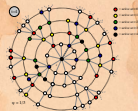
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{\# outgoing infected edges}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

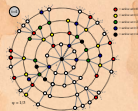
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$R = \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}}$$

$$+ \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{\# outgoing infected edges}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}$$

Contagion

Basic Contagion Models

Global spreading co

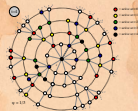
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Global spreading condition

- ▶ We need to find: <sup>[5]</sup>

**R** = the average # of infected edges that one random infected edge brings about.

- ▶ Call **R** the **gain ratio**.

- ▶ Define  $B_{k1}$  as the probability that a node of degree  $k$  is infected by a single infected edge.

- ▶

$$\begin{aligned}
 R = & \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{prob. of connecting to a degree } k \text{ node}} \cdot \underbrace{(k-1)}_{\text{\# outgoing infected edges}} \cdot \underbrace{B_{k1}}_{\text{Prob. of infection}} \\
 & + \sum_{k=0}^{\infty} \underbrace{\frac{kP_k}{\langle k \rangle}}_{\text{\# outgoing infected edges}} \cdot \underbrace{0}_{\text{\# outgoing infected edges}} \cdot \underbrace{(1 - B_{k1})}_{\text{Prob. of no infection}}
 \end{aligned}$$

Contagion

Basic Contagion Models

Global spreading co

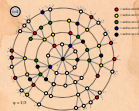
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ Our global spreading condition is then:

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} > 1.$$

- ▶ Case 1: If  $B_{k1} = 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) = \frac{\langle k(k-1) \rangle}{\langle k \rangle} > 1.$$

- ▶ Good: This is just our giant component condition again.

Contagion

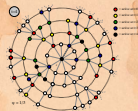
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ Our global spreading condition is then:

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} > 1.$$

- ▶ **Case 1:** If  $B_{k1} = 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) = \frac{\langle k(k-1) \rangle}{\langle k \rangle} > 1.$$

- ▶ **Good:** This is just our giant component condition again.

Contagion

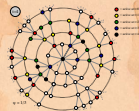
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ Our global spreading condition is then:

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} > 1.$$

- ▶ **Case 1:** If  $B_{k1} = 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) = \frac{\langle k(k-1) \rangle}{\langle k \rangle} > 1.$$

- ▶ **Good:** This is just our giant component condition again.

Contagion

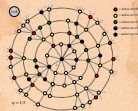
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ Our global spreading condition is then:

$$\mathbf{R} = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} > 1.$$

- ▶ **Case 1:** If  $B_{k1} = 1$  then

$$\mathbf{R} = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) = \frac{\langle k(k-1) \rangle}{\langle k \rangle} > 1.$$

- ▶ Good: This is just our giant component condition again.

Contagion

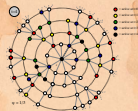
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ Our global spreading condition is then:

$$\mathbf{R} = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} > 1.$$

- ▶ **Case 1:** If  $B_{k1} = 1$  then

$$\mathbf{R} = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) = \frac{\langle k(k-1) \rangle}{\langle k \rangle} > 1.$$

- ▶ **Good:** This is just our giant component condition again.

Contagion

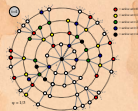
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) \bullet \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but critical value of  $\langle k \rangle$  is increased.
- ▶ Aka bond percolation ( $\boxplus$ ).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 ( $\boxplus$ )

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

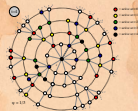
Basic Contagion Models

Global spreading co

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) \bullet \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but critical value of  $\langle k \rangle$  is increased.
- ▶ Aka bond percolation ( $\boxplus$ ).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 ( $\boxplus$ )

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

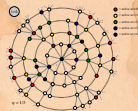
Basic Contagion Models

Global spreading co

Social Contagion Models

Network version  
All-to-all networks  
Theory

References





# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but critical value of  $\langle k \rangle$  is increased.
- ▶ Aka bond percolation ( $\boxplus$ ).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 ( $\boxplus$ )

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

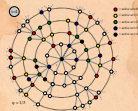
Basic Contagion Models

Global spreading co

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but critical value of  $\langle k \rangle$  is increased.
- ▶ Aka bond percolation ( $\boxplus$ ).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 ( $\boxplus$ )

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

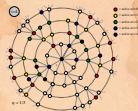
Basic Contagion Models

Global spreading co

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but **critical value** of  $\langle k \rangle$  is **increased**.
- ▶ Aka bond percolation ( $\boxplus$ ).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 ( $\boxplus$ )

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

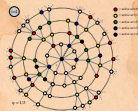
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but **critical value** of  $\langle k \rangle$  is **increased**.
- ▶ Aka bond percolation (⊕).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 (⊕)

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

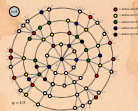
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but **critical value** of  $\langle k \rangle$  is **increased**.
- ▶ Aka bond percolation (田).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 (田)

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

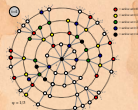
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Case 2:** If  $B_{k1} = \beta < 1$  then

$$R = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot \beta > 1.$$

- ▶ A fraction  $(1-\beta)$  of edges do not transmit infection.
- ▶ Analogous phase transition to giant component case but **critical value** of  $\langle k \rangle$  is **increased**.
- ▶ Aka bond percolation (田).
- ▶ Resulting degree distribution  $P'_k$ :

$$P'_k = \beta^k \sum_{i=k}^{\infty} \binom{i}{k} (1-\beta)^{i-k} P_i.$$

Insert question from assignment 7 (田)

- ▶ We can show  $F_{P'}(x) = F_P(\beta x + 1 - \beta)$ .

Contagion

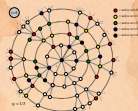
Basic Contagion  
Models

Global spreading co

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k,1}$  to depend on  $k$
- ▶ Asymmetry: Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k,1}$  increases with  $k$ ... unlikely.
- ▶ Possibility:  $B_{k,1}$  is not monotonic in  $k$ ... unlikely.
- ▶ Possibility:  $B_{k,1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k,1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story: More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

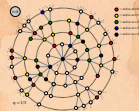
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ Asymmetry: Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... unlikely.
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... unlikely.
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story: More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

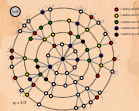
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... unlikely
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... unlikely.
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story:  
More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

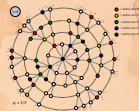
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ **Possibility:**  $B_{k1}$  increases with  $k$ ... unlikely.
- ▶ **Possibility:**  $B_{k1}$  is not monotonic in  $k$ ... unlikely.
- ▶ **Possibility:**  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ **The story:** More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

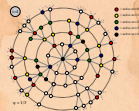
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ **Possibility:**  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ **Possibility:**  $B_{k1}$  is not monotonic in  $k$ ... unlikely.
- ▶ **Possibility:**  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ **The story:** More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

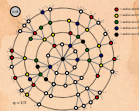
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story:  
More well connected people are harder to influence.

Contagion

Basic Contagion  
Models

Global spreading  
condition

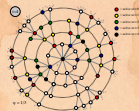
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... hmmm.
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story:  
More well connected people are harder to influence.

Contagion

Basic Contagion  
Models

Global spreading  
condition

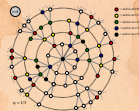
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... **hmmm.**
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ The story:  
More well connected people are harder to influence.

Contagion

Basic Contagion Models

Global spreading condition

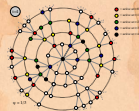
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... **hmmm.**
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ **The story:**  
More well connected people are harder to influence.

Contagion

Basic Contagion  
Models

Global spreading  
condition

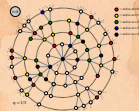
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Cases 3, 4, 5, ...:** Now allow  $B_{k1}$  to depend on  $k$
- ▶ **Asymmetry:** Transmission along an edge depends on node's degree at other end.
- ▶ Possibility:  $B_{k1}$  increases with  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  is not monotonic in  $k$ ... **unlikely.**
- ▶ Possibility:  $B_{k1}$  decreases with  $k$ ... **hmmm.**
- ▶  $B_{k1} \searrow$  is a plausible representation of a simple kind of social contagion.
- ▶ **The story:**  
More well connected people are harder to influence.

Contagion

Basic Contagion  
Models

Global spreading  
condition

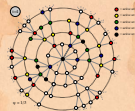
Social Contagion  
Models

Network version

All-to-all networks

Theory

References





# Global spreading condition

- ▶ **Example:**  $B_{k1} = 1/k$ .

$$\begin{aligned} R &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k} \\ &= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle} \end{aligned}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

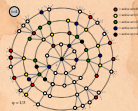
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = 1/k$ .



$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k}$$

$$= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

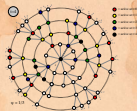
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

▶ **Example:**  $B_{k1} = 1/k$ .

▶

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k}$$

$$= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

Basic Contagion  
Models

Global spreading  
condition

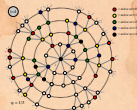
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

▶ **Example:**  $B_{k1} = 1/k$ .

▶

$$\begin{aligned} R &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k} \\ &= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle} \end{aligned}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

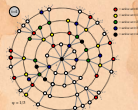
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

▶ **Example:**  $B_{k1} = 1/k$ .

▶

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k}$$

$$= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

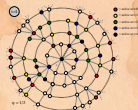
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

▶ **Example:**  $B_{k1} = 1/k$ .

▶

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k}$$

$$= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle}$$

- ▶ Since  $R$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

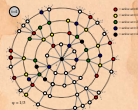
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Global spreading condition

▶ **Example:**  $B_{k1} = 1/k$ .

▶

$$\begin{aligned} \mathbf{R} &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k} \\ &= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle} \end{aligned}$$

- ▶ Since  $\mathbf{R}$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

Basic Contagion  
Models

Global spreading  
condition

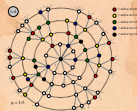
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = 1/k$ .



$$\begin{aligned} \mathbf{R} &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k} \\ &= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle} \end{aligned}$$

- ▶ Since  $\mathbf{R}$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

Basic Contagion  
Models

Global spreading  
condition

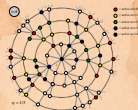
Social Contagion  
Models

Network version

All-to-all networks

Theory

References





# Global spreading condition

- ▶ **Example:**  $B_{k1} = 1/k$ .



$$\begin{aligned} \mathbf{R} &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \cdot \frac{1}{k} \\ &= \sum_{k=1}^{\infty} \frac{P_k}{\langle k \rangle} \cdot (k-1) = 1 - \frac{1 - P_0}{\langle k \rangle} \end{aligned}$$

- ▶ Since  $\mathbf{R}$  is always less than 1, no spreading can occur for this mechanism.
- ▶ Decay of  $B_{k1}$  is too fast.
- ▶ Result is independent of degree distribution.

Contagion

Basic Contagion  
Models

Global spreading  
condition

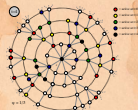
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).

- ▶ Infection only occurs for nodes with low degree.
- ▶ Call these nodes **vulnerables**:  
they flip when only one of their friends flips.
- ▶

$$\begin{aligned} R &= \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right) \\ &= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.} \end{aligned}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

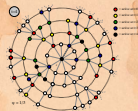
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).
- ▶ Infection only occurs for nodes with **low** degree.
- ▶ Call these nodes **vulnerables**:  
they flip when only one of their friends flips.

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right)$$
$$= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

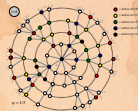
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).
- ▶ Infection only occurs for nodes with **low** degree.
- ▶ Call these nodes **vulnerables**:  
they flip when **only one** of their friends flips.

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right)$$
$$= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

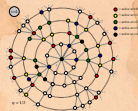
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).
- ▶ Infection only occurs for nodes with **low** degree.
- ▶ Call these nodes **vulnerables**:  
they flip when **only one** of their friends flips.
- ▶

$$\mathbf{R} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right)$$
$$= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

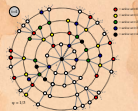
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).
- ▶ Infection only occurs for nodes with **low** degree.
- ▶ Call these nodes **vulnerables**:  
they flip when **only one** of their friends flips.
- ▶

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right)$$

$$= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

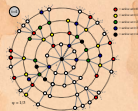
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ **Example:**  $B_{k1} = H\left(\frac{1}{k} - \phi\right)$   
where  $0 < \phi \leq 1$  is a **threshold** and  $H$  is the Heaviside function ( $\boxplus$ ).
- ▶ Infection only occurs for nodes with **low** degree.
- ▶ Call these nodes **vulnerables**:  
they flip when **only one** of their friends flips.
- ▶

$$\mathbf{R} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot B_{k1} = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \cdot (k-1) \cdot H\left(\frac{1}{k} - \phi\right)$$
$$= \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} \quad \text{where } \lfloor \cdot \rfloor \text{ means floor.}$$

Contagion

Basic Contagion  
Models

Global spreading  
condition

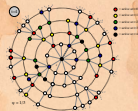
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ The uniform threshold model global spreading condition:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} > 1.$$

- ▶ As  $\phi \rightarrow 1$ , all nodes become resilient and  $r \rightarrow 0$ .
- ▶ As  $\phi \rightarrow 0$ , all nodes become vulnerable and the contagion condition matches up with the giant component condition.
- ▶ Key: If we fix  $\phi$  and then vary  $\langle k \rangle$ , we may see two phase transitions.
- ▶ Added to our standard giant component transition, we will see a cut off in spreading as nodes become more connected.

Contagion

Basic Contagion Models

Global spreading condition

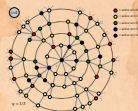
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Global spreading condition

- ▶ The uniform threshold model global spreading condition:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} > 1.$$

- ▶ As  $\phi \rightarrow 1$ , all nodes become resilient and  $r \rightarrow 0$ .
- ▶ As  $\phi \rightarrow 0$ , all nodes become vulnerable and the contagion condition matches up with the giant component condition.
- ▶ Key: If we fix  $\phi$  and then vary  $\langle k \rangle$ , we may see two phase transitions.
- ▶ Added to our standard giant component transition, we will see a cut off in spreading as nodes become more connected.

Contagion

Basic Contagion Models

Global spreading condition

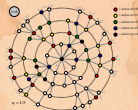
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ The uniform threshold model global spreading condition:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} > 1.$$

- ▶ As  $\phi \rightarrow 1$ , all nodes become resilient and  $r \rightarrow 0$ .
- ▶ As  $\phi \rightarrow 0$ , all nodes become vulnerable and the contagion condition matches up with the giant component condition.
- ▶ Key: If we fix  $\phi$  and then vary  $\langle k \rangle$ , we may see two phase transitions.
- ▶ Added to our standard giant component transition, we will see a cut off in spreading as nodes become more connected.

Contagion

Basic Contagion Models

Global spreading condition

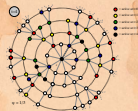
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ The uniform threshold model global spreading condition:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} > 1.$$

- ▶ As  $\phi \rightarrow 1$ , all nodes become resilient and  $r \rightarrow 0$ .
- ▶ As  $\phi \rightarrow 0$ , all nodes become vulnerable and the contagion condition matches up with the giant component condition.
- ▶ **Key:** If we fix  $\phi$  and then vary  $\langle k \rangle$ , we may see **two** phase transitions.
- ▶ Added to our standard giant component transition, we will see a cut off in spreading as nodes become more connected.

Contagion

Basic Contagion  
Models

Global spreading  
condition

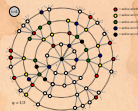
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Global spreading condition

- ▶ The uniform threshold model global spreading condition:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} (k-1) \cdot \frac{kP_k}{\langle k \rangle} > 1.$$

- ▶ As  $\phi \rightarrow 1$ , all nodes become resilient and  $r \rightarrow 0$ .
- ▶ As  $\phi \rightarrow 0$ , all nodes become vulnerable and the contagion condition matches up with the giant component condition.
- ▶ **Key:** If we fix  $\phi$  and then vary  $\langle k \rangle$ , we may see **two** phase transitions.
- ▶ Added to our standard giant component transition, we will see a cut off in spreading as nodes become more connected.

Contagion

Basic Contagion Models

Global spreading condition

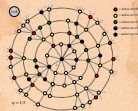
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Outline

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

All-to-all networks

Theory

References

Contagion

Basic Contagion  
Models

Global spreading  
condition

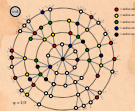
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory, Informational cascades,...

Contagion

Basic Contagion Models

Global spreading condition

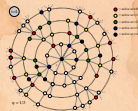
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory, Informational cascades,...

Contagion

Basic Contagion Models

Global spreading condition

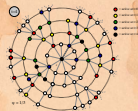
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory, Informational cascades,...

Contagion

Basic Contagion Models

Global spreading condition

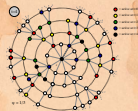
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory, Informational cascades,...

Contagion

Basic Contagion  
Models

Global spreading  
condition

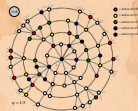
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory: Informational cascades,...

Contagion

Basic Contagion Models

Global spreading condition

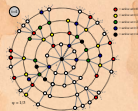
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Social Contagion

## Some important models (recap from CSYS 300)

- ▶ Tipping models—Schelling (1971) [9, 10, 11]
  - ▶ Simulation on checker boards.
  - ▶ Idea of thresholds.
- ▶ Threshold models—Granovetter (1978) [8]
- ▶ Herding models—Bikhchandani et al. (1992) [1, 2]
  - ▶ Social learning theory, Informational cascades,...

Contagion

Basic Contagion  
Models

Global spreading  
condition

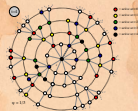
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold model on a network

Original work:

**“A simple model of global cascades on random networks”**

D. J. Watts. Proc. Natl. Acad. Sci., 2002<sup>[13]</sup>

- ▶ Mean field Granovetter model  $\rightarrow$  network model
- ▶ Individuals now have a limited view of the world

Contagion

Basic Contagion  
Models

Global spreading  
condition

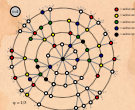
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold model on a network

Original work:

**“A simple model of global cascades on random networks”**

D. J. Watts. Proc. Natl. Acad. Sci., 2002<sup>[13]</sup>

- ▶ Mean field Granovetter model → network model
- ▶ Individuals now have a limited view of the world

Contagion

Basic Contagion  
Models

Global spreading  
condition

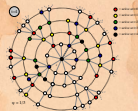
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold model on a network

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks

Theory

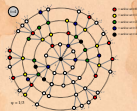
References

Original work:

**“A simple model of global cascades on random networks”**

D. J. Watts. Proc. Natl. Acad. Sci., 2002 <sup>[13]</sup>

- ▶ Mean field Granovetter model → network model
- ▶ Individuals now have a limited view of the world



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is sparse
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is reciprocal and of unit weight
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

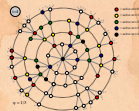
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is reciprocal and of unit weight
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

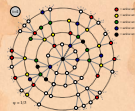
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References





# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is reciprocal and of unit weight
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

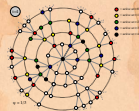
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

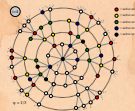
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

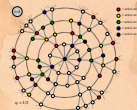
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous: discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

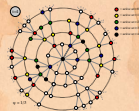
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous, discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i/k_i$
- ▶ Activation is permanent (SI)

Contagion

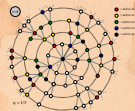
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous, discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i k_i$
- ▶ Activation is permanent (SI)

Contagion

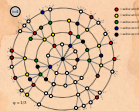
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network

- ▶ Interactions between individuals now represented by a network
- ▶ Network is **sparse**
- ▶ Individual  $i$  has  $k_i$  contacts
- ▶ Influence on each link is **reciprocal** and of **unit weight**
- ▶ Each individual  $i$  has a fixed threshold  $\phi_i$
- ▶ Individuals repeatedly poll contacts on network
- ▶ Synchronous, discrete time updating
- ▶ Individual  $i$  becomes active when number of active contacts  $a_i \geq \phi_i k_i$
- ▶ Activation is permanent (SI)

Contagion

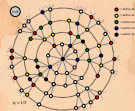
Basic Contagion Models

Global spreading condition

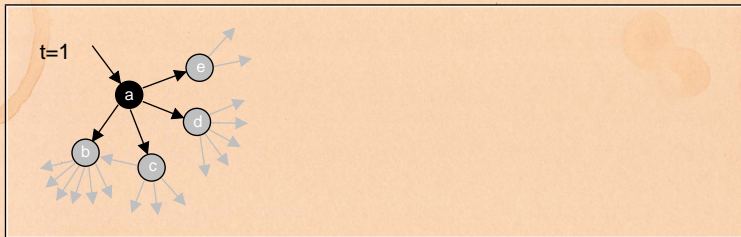
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Threshold model on a network



▶ All nodes have threshold  $\phi = 0.2$ .

Contagion

Basic Contagion Models

Global spreading condition

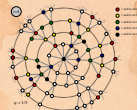
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Threshold model on a network

Contagion

Basic Contagion Models

Global spreading condition

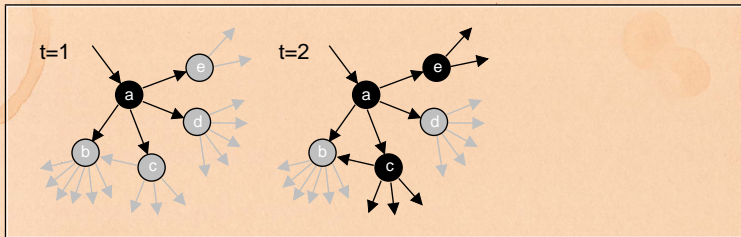
Social Contagion Models

Network version

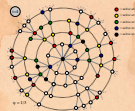
All-to-all networks

Theory

References



- ▶ All nodes have threshold  $\phi = 0.2$ .



# Threshold model on a network

Contagion

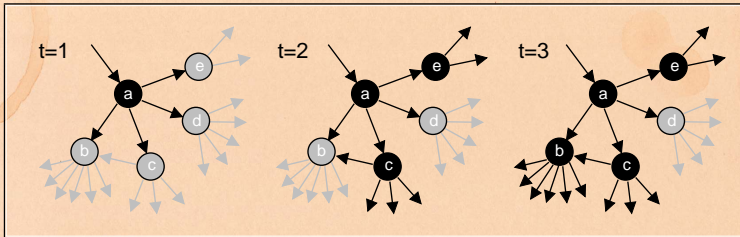
Basic Contagion Models

Global spreading condition

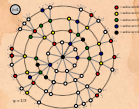
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



- ▶ All nodes have threshold  $\phi = 0.2$ .



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ **Key:** For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

Basic Contagion Models

Global spreading condition

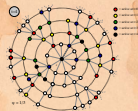
Social Contagion Models

Network version

All-to-all networks

Theory

References



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ Key: For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

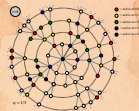
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ Key: For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

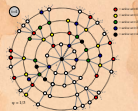
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ Key: For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

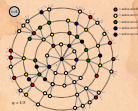
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ **Key:** For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

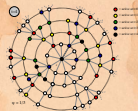
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# The most gullible

## Vulnerables:

- ▶ Recall definition: individuals who can be activated by just one contact being active are **vulnerables**.
- ▶ The vulnerability condition for node  $i$ :  $1/k_i \geq \phi_i$ .
- ▶ Means # contacts  $k_i \leq \lfloor 1/\phi_i \rfloor$ .
- ▶ **Key:** For global spreading events (cascades) on random networks, must have a *global component of vulnerables*<sup>[13]</sup>
- ▶ For a uniform threshold  $\phi$ , our global spreading condition tells us when such a component exists:

$$R = \sum_{k=1}^{\lfloor \frac{1}{\phi} \rfloor} \frac{kP_k}{\langle k \rangle} \bullet (k-1) > 1.$$

Contagion

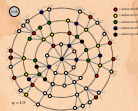
Basic Contagion Models

Global spreading condition

Social Contagion Models

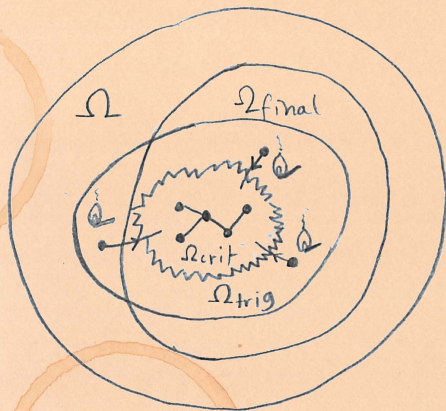
Network version  
All-to-all networks  
Theory

References





# Example random network structure:



- ▶  $\Omega_{crit}$  = critical mass = global vulnerable component
- ▶  $\Omega_{trig}$  = triggering component
- ▶  $\Omega_{final}$  = potential extent of spread
- ▶  $\Omega$  = entire network

$$\Omega_{crit} \subset \Omega_{trig}; \quad \Omega_{crit} \subset \Omega_{final}; \quad \text{and} \quad \Omega_{trig}, \Omega_{final} \subset \Omega.$$

Contagion

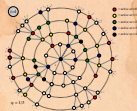
Basic Contagion Models

Global spreading condition

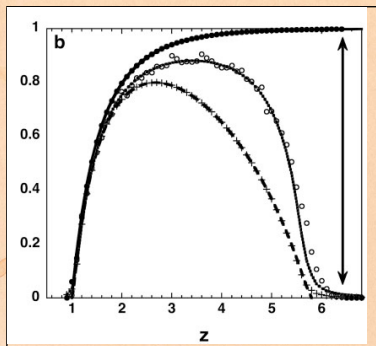
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



(n.b.,  $z = \langle k \rangle$ )

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

Contagion

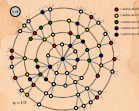
Basic Contagion Models

Global spreading condition

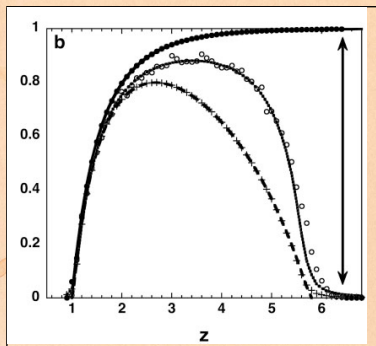
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



( n.b.,  $z = \langle k \rangle$  )

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

Contagion

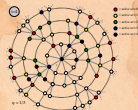
Basic Contagion Models

Global spreading condition

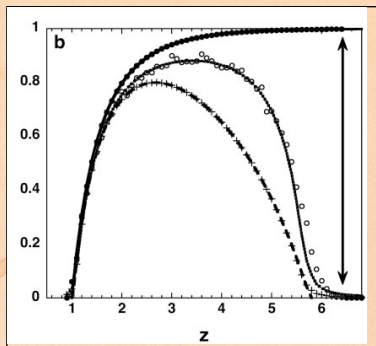
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



( n.b.,  $z = \langle k \rangle$  )

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

Contagion

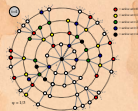
Basic Contagion Models

Global spreading condition

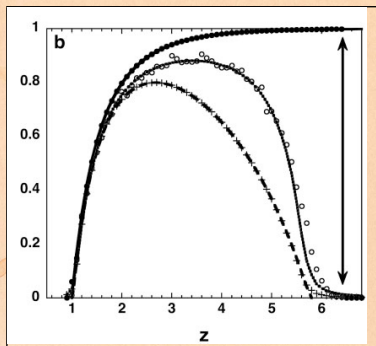
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



(n.b.,  $z = \langle k \rangle$ )

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

Contagion

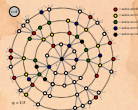
Basic Contagion Models

Global spreading condition

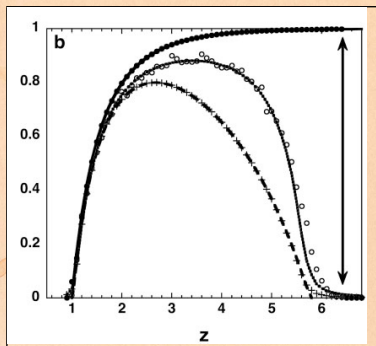
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



(n.b.,  $z = \langle k \rangle$ )

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

Contagion

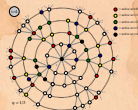
Basic Contagion Models

Global spreading condition

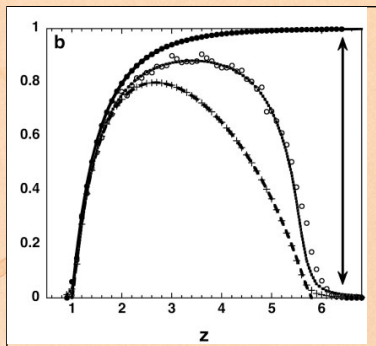
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Global spreading events on random networks



(n.b.,  $z = \langle k \rangle$ )

- ▶ Global spreading events occur only if size of vulnerable subcomponent  $> 0$ .
- ▶ System is robust-yet-fragile just below upper boundary [3, 4, 12]
- ▶ 'Ignorance' facilitates spreading.

- ▶ **Top curve:** final fraction infected if successful.
- ▶ **Middle curve:** chance of starting a global spreading event (cascade).
- ▶ **Bottom curve:** fractional size of vulnerable subcomponent. [13]

Contagion

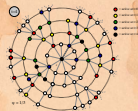
Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Cascades on random networks

Contagion

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

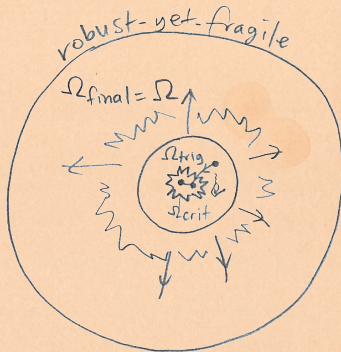
All-to-all networks

Theory

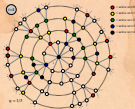
References



- ▶ Above lower phase transition

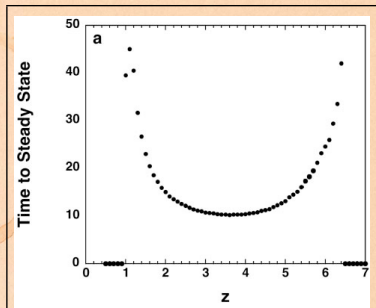


- ▶ Just below upper phase transition





# Cascades on random networks



( n.b.,  $z = \langle k \rangle$  )

- ▶ Largest vulnerable component = critical mass.
- ▶ Now have endogenous mechanism for spreading from an individual to the critical mass and then beyond.

- ▶ Time taken for cascade to spread through network. [13]
- ▶ Two phase transitions.

Contagion

Basic Contagion Models

Global spreading condition

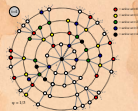
Social Contagion Models

Network version

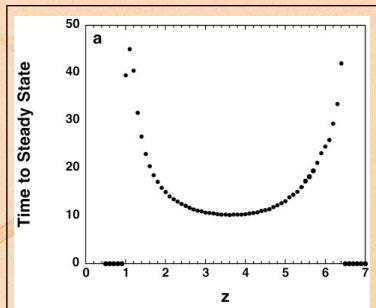
All-to-all networks

Theory

References



# Cascades on random networks



( n.b.,  $z = \langle k \rangle$  )

- ▶ Largest vulnerable component = critical mass.
- ▶ Now have endogenous mechanism for spreading from an individual to the critical mass and then beyond.

- ▶ Time taken for cascade to spread through network. [13]
- ▶ Two phase transitions.

Contagion

Basic Contagion Models

Global spreading condition

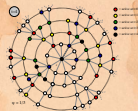
Social Contagion Models

Network version

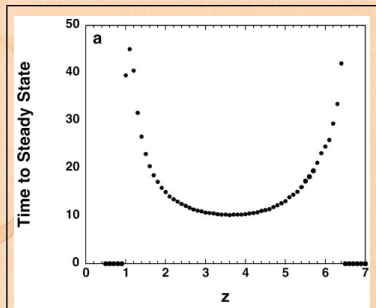
All-to-all networks

Theory

References



# Cascades on random networks



( n.b.,  $z = \langle k \rangle$  )

- ▶ Largest vulnerable component = **critical mass**.
- ▶ Now have endogenous mechanism for spreading from an individual to the critical mass and then beyond.

- ▶ Time taken for cascade to spread through network. [13]
- ▶ Two phase transitions.

Contagion

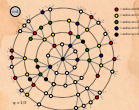
Basic Contagion Models

Global spreading condition

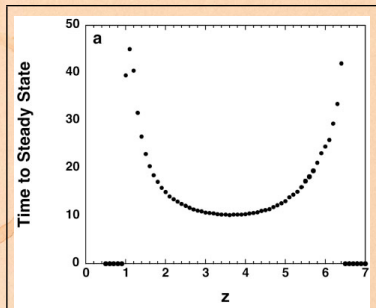
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Cascades on random networks



( n.b.,  $z = \langle k \rangle$ )

- ▶ Largest vulnerable component = **critical mass**.
- ▶ Now have endogenous mechanism for spreading from an individual to the critical mass and then beyond.

- ▶ Time taken for cascade to spread through network. [13]
- ▶ Two phase transitions.

Contagion

Basic Contagion Models

Global spreading condition

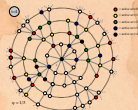
Social Contagion Models

Network version

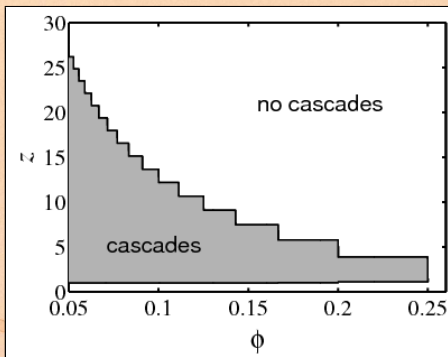
All-to-all networks

Theory

References



# Cascade window for random networks



(n.b.,  $z = \langle k \rangle$ )

- ▶ Outline of cascade window for random networks.

Contagion

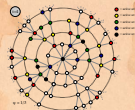
Basic Contagion  
Models

Global spreading  
condition

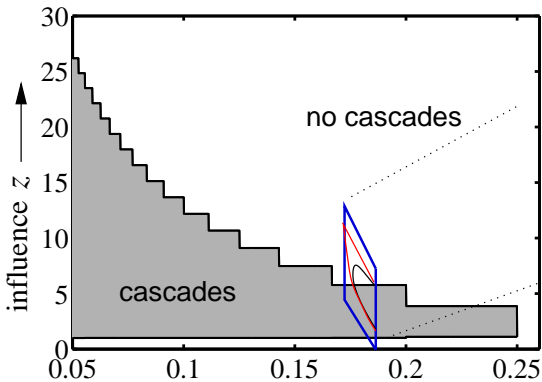
Social Contagion  
Models

Network version  
All-to-all networks  
Theory

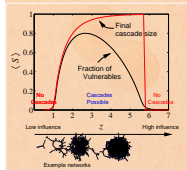
References



# Cascade window for random networks



$\phi$  = uniform individual threshold



Contagion

Basic Contagion Models

Global spreading condition

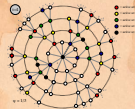
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Outline

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

All-to-all networks

Theory

References

Contagion

Basic Contagion  
Models

Global spreading  
condition

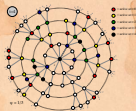
Social Contagion  
Models

Network version

All-to-all networks

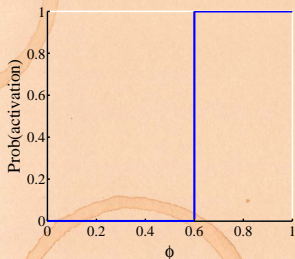
Theory

References



# Social Contagion

## Granovetter's Threshold model—recap



- ▶ Assumes deterministic response functions

- ▶  $\phi_c =$  threshold of an individual.

- ▶  $f(\phi_c) =$  distribution of thresholds in a population.

- ▶  $F(\phi_c) =$  cumulative distribution =  $\int_{\phi_c}^1 f(\phi_c) d\phi_c$

- ▶  $\phi_t =$  fraction of people 'rioting' at time step  $t$ .

Contagion

Basic Contagion Models

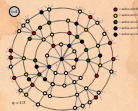
Global spreading condition

Social Contagion Models

Network version

All-to-all networks Theory

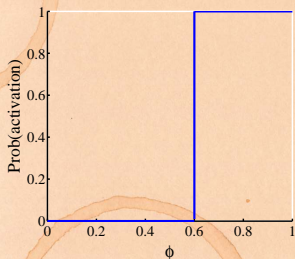
References





# Social Contagion

## Granovetter's Threshold model—recap



- ▶ Assumes deterministic response functions
- ▶  $\phi_*$  = threshold of an individual.
- ▶  $f(\phi_*)$  = distribution of thresholds in a population.
- ▶  $F(\phi_*)$  = cumulative distribution =  $\int_{\phi_*}^1 f(\phi'_*) d\phi'_*$
- ▶  $\phi_t$  = fraction of people 'rioting' at time step  $t$ .

Contagion

Basic Contagion Models

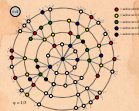
Global spreading condition

Social Contagion Models

Network version

All-to-all networks  
Theory

References



# Social Contagion

Contagion

Basic Contagion Models

Global spreading condition

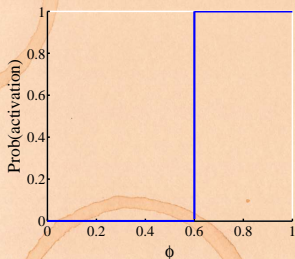
Social Contagion Models

Network version

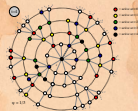
All-to-all networks Theory

References

## Granovetter's Threshold model—recap



- ▶ Assumes deterministic response functions
- ▶  $\phi_*$  = threshold of an individual.
- ▶  $f(\phi_*)$  = distribution of thresholds in a population.
- ▶  $F(\phi_*)$  = cumulative distribution =  $\int_{\phi_*}^1 f(\phi_*) d\phi_*$
- ▶  $\phi_t$  = fraction of people 'rioting' at time step  $t$ .



# Social Contagion

Contagion

Basic Contagion Models

Global spreading condition

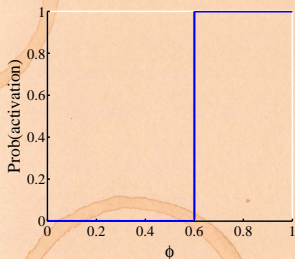
Social Contagion Models

Network version

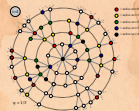
All-to-all networks Theory

References

## Granovetter's Threshold model—recap



- ▶ Assumes deterministic response functions
- ▶  $\phi_*$  = threshold of an individual.
- ▶  $f(\phi_*)$  = distribution of thresholds in a population.
- ▶  $F(\phi_*)$  = cumulative distribution =  $\int_{\phi'_*=0}^{\phi_*} f(\phi'_*)d\phi'_*$
- ▶  $\phi_t$  = fraction of people 'rioting' at time step  $t$ .



# Social Contagion

Contagion

Basic Contagion Models

Global spreading condition

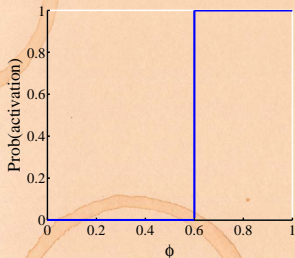
Social Contagion Models

Network version

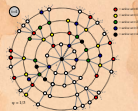
All-to-all networks Theory

References

## Granovetter's Threshold model—recap



- ▶ Assumes deterministic response functions
- ▶  $\phi_*$  = threshold of an individual.
- ▶  $f(\phi_*)$  = distribution of thresholds in a population.
- ▶  $F(\phi_*)$  = cumulative distribution =  $\int_{\phi'_*=0}^{\phi_*} f(\phi'_*)d\phi'_*$
- ▶  $\phi_t$  = fraction of people 'rioting' at time step  $t$ .



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

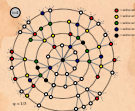
All-to-all networks  
Theory

References

- ▶ At time  $t + 1$ , fraction rioting = fraction with  $\phi_* \leq \phi_t$ .

$$\phi_{t+1} = \int_0^{\phi_t} f(\phi_*) d\phi_* = F(\phi_*) \Big|_0^{\phi_t} = F(\phi_t)$$

- ▶  $\Rightarrow$  Iterative maps of the unit interval  $[0, 1]$ .



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks

Theory

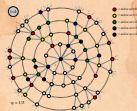
References

▶ At time  $t + 1$ , fraction rioting = fraction with  $\phi_* \leq \phi_t$ .



$$\phi_{t+1} = \int_0^{\phi_t} f(\phi_*) d\phi_* = F(\phi_*)|_0^{\phi_t} = F(\phi_t)$$

▶  $\Rightarrow$  Iterative maps of the unit interval  $[0, 1]$ .



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks

Theory

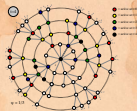
References

- ▶ At time  $t + 1$ , fraction rioting = fraction with  $\phi_* \leq \phi_t$ .



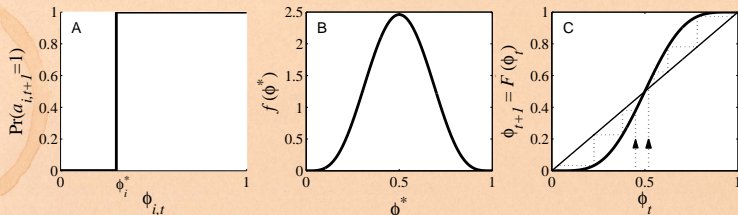
$$\phi_{t+1} = \int_0^{\phi_t} f(\phi_*) d\phi_* = F(\phi_*)|_0^{\phi_t} = F(\phi_t)$$

- ▶  $\Rightarrow$  Iterative maps of the unit interval  $[0, 1]$ .



# Social Sciences—Threshold models

Action based on perceived behavior of others.



- ▶ Two states: S and I
- ▶ Recover now possible (SIS)
- ▶  $\phi$  = fraction of contacts 'on' (e.g., rioting)
- ▶ Discrete time, synchronous update (strong assumption!)
- ▶ This is a Critical mass model

Contagion

Basic Contagion Models

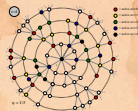
Global spreading condition

Social Contagion Models

Network version

All-to-all networks Theory

References





# Social Sciences—Threshold models

Contagion

Basic Contagion Models

Global spreading condition

Social Contagion Models

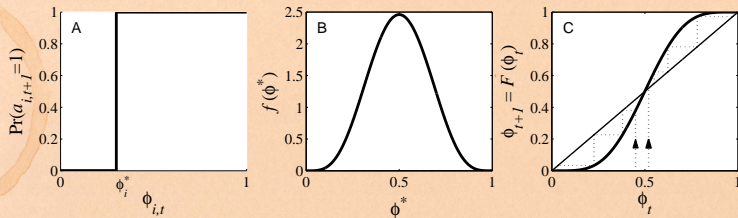
Network version

All-to-all networks

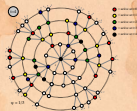
Theory

References

Action based on perceived behavior of others.



- ▶ Two states: S and I
- ▶ Recover now possible (SIS)
- ▶  $\phi$  = fraction of contacts 'on' (e.g., rioting)
- ▶ Discrete time, synchronous update (strong assumption!)
- ▶ This is a Critical mass model



# Social Sciences—Threshold models

Contagion

Basic Contagion Models

Global spreading condition

Social Contagion Models

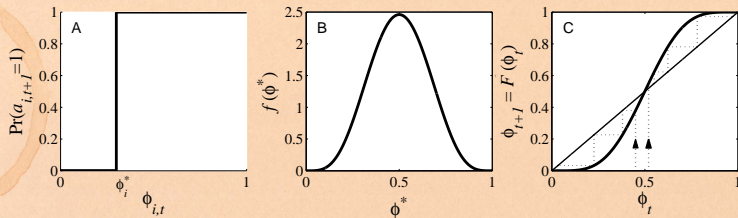
Network version

All-to-all networks

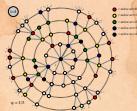
Theory

References

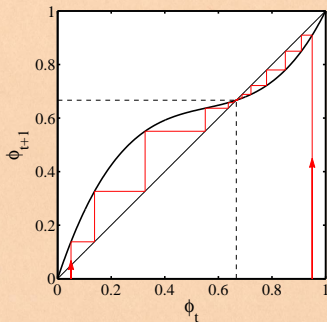
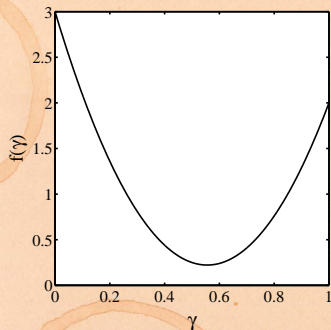
Action based on perceived behavior of others.



- ▶ Two states: S and I
- ▶ Recover now possible (SIS)
- ▶  $\phi$  = fraction of contacts 'on' (e.g., rioting)
- ▶ Discrete time, synchronous update (strong assumption!)
- ▶ This is a **Critical mass model**



# Social Sciences—Threshold models



- ▶ Example of single stable state model

Contagion

Basic Contagion Models

Global spreading condition

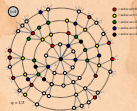
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Social Sciences—Threshold models

## Implications for collective action theory:

1. Collective uniformity  $\nrightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.

Contagion

Basic Contagion  
Models

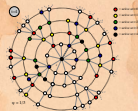
Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

References



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

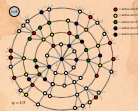
References

## Implications for collective action theory:

1. Collective uniformity  $\not\Rightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

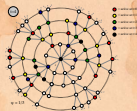
References

## Implications for collective action theory:

1. Collective uniformity  $\not\Rightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

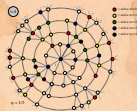
References

## Implications for collective action theory:

1. Collective uniformity  $\not\Rightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.



# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

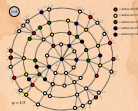
References

## Implications for collective action theory:

1. Collective uniformity  $\not\Rightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.





# Social Sciences—Threshold models

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks  
Theory

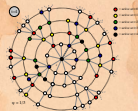
References

## Implications for collective action theory:

1. Collective uniformity  $\not\Rightarrow$  individual uniformity
2. Small individual changes  $\Rightarrow$  large global changes

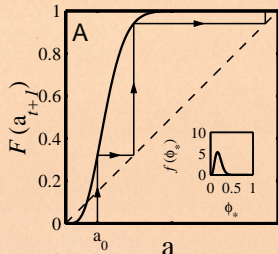
## Next:

- ▶ Connect mean-field model to network model.
- ▶ Single seed for network model:  $1/N \rightarrow 0$ .
- ▶ Comparison between network and mean-field model sensible for vanishing seed size for the latter.

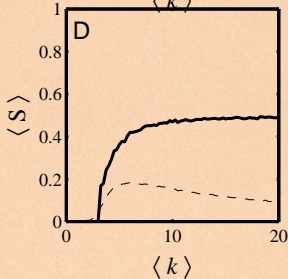
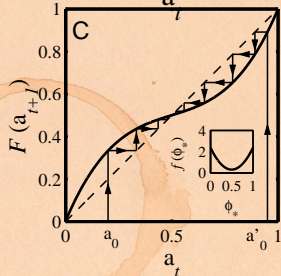
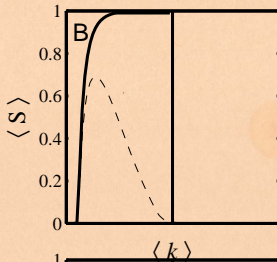


# All-to-all versus random networks

all-to-all networks



random networks



Contagion

Basic Contagion Models

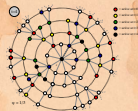
Global spreading condition

Social Contagion Models

Network version

All-to-all networks Theory

References



# Outline

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

All-to-all networks

Theory

References

Contagion

Basic Contagion  
Models

Global spreading  
condition

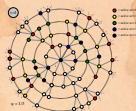
Social Contagion  
Models

Network version

All-to-all networks

**Theory**

References



# Threshold contagion on random networks

## Three key pieces to describe analytically:

1. The fractional size of the largest subcomponent of vulnerable nodes,  $S_{\text{vuln}}$ .
2. The chance of starting a global spreading event,  $P_{\text{trig}} = S_{\text{trig}}$ .
3. The expected final size of any successful spread,  $S$ .
  - ▶ n.b., the distribution of  $S$  is almost always bimodal.

Contagion

Basic Contagion Models

Global spreading condition

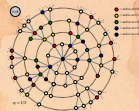
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

## Three key pieces to describe analytically:

1. The fractional size of the largest subcomponent of vulnerable nodes,  $S_{\text{vuln}}$ .
2. The chance of starting a global spreading event,  $P_{\text{trig}} = S_{\text{trig}}$ .
3. The expected final size of any successful spread,  $S$ .
  - ▶ n.b., the distribution of  $S$  is almost always bimodal.

Contagion

Basic Contagion Models

Global spreading condition

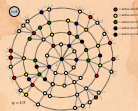
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

## Three key pieces to describe analytically:

1. The fractional size of the largest subcomponent of vulnerable nodes,  $S_{\text{vuln}}$ .
2. The chance of starting a global spreading event,  $P_{\text{trig}} = S_{\text{trig}}$ .
3. The expected final size of any successful spread,  $S$ .
  - ▶ n.b., the distribution of  $S$  is almost always bimodal.

Contagion

Basic Contagion Models

Global spreading condition

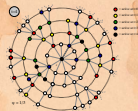
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

## Three key pieces to describe analytically:

1. The fractional size of the largest subcomponent of vulnerable nodes,  $S_{\text{vuln}}$ .
2. The chance of starting a global spreading event,  $P_{\text{trig}} = S_{\text{trig}}$ .
3. The expected final size of any successful spread,  $S$ .
  - ▶ n.b., the distribution of  $S$  is almost always bimodal.

Contagion

Basic Contagion Models

Global spreading condition

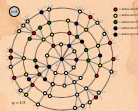
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

## Three key pieces to describe analytically:

1. The fractional size of the largest subcomponent of vulnerable nodes,  $S_{\text{vuln}}$ .
2. The chance of starting a global spreading event,  $P_{\text{trig}} = S_{\text{trig}}$ .
3. The expected final size of any successful spread,  $S$ .
  - ▶ n.b., the distribution of  $S$  is almost always bimodal.

Contagion

Basic Contagion Models

Global spreading condition

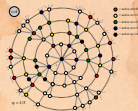
Social Contagion Models

Network version

All-to-all networks

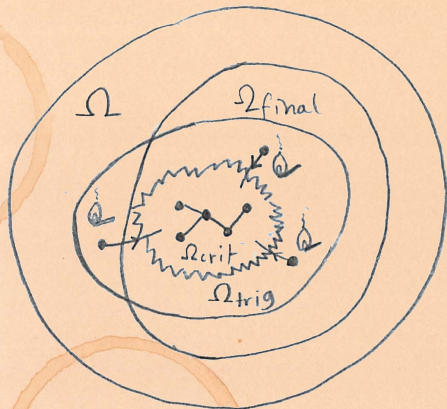
Theory

References





# Example random network structure:



- ▶  $\Omega_{\text{crit}} = \Omega_{\text{vuln}} =$   
critical mass =  
global  
vulnerable  
component
- ▶  $\Omega_{\text{trig}} =$   
triggering  
component
- ▶  $\Omega_{\text{final}} =$   
potential extent  
of spread
- ▶  $\Omega =$  entire  
network

$$\Omega_{\text{crit}} \subset \Omega_{\text{trig}}; \Omega_{\text{crit}} \subset \Omega_{\text{final}}; \text{ and } \Omega_{\text{trig}}, \Omega_{\text{final}} \subset \Omega.$$

Contagion

Basic Contagion  
Models

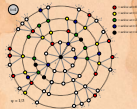
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **First goal:** Find the largest component of vulnerable nodes.

- ▶ Recall that for finding the giant component's size, we had to solve:

$$F_v(x) = xF_P(F_v(x)) \quad \text{and} \quad F_u(x) = xF_R(F_v(x))$$

- ▶ We'll find a similar result for the subset of nodes that are vulnerable.
- ▶ This is a node-based percolation problem.
- ▶ For a general monotonic threshold distribution  $f(\phi)$ , a degree  $k$  node is vulnerable with probability

$$B_{k\gamma} = \int_0^{1/k} f(\phi) d\phi.$$

Contagion

Basic Contagion Models

Global spreading condition

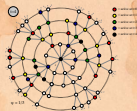
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **First goal:** Find the largest component of vulnerable nodes.
- ▶ Recall that for finding the giant component's size, we had to solve:

$$F_{\pi}(x) = xF_P(F_{\rho}(x)) \quad \text{and} \quad F_{\rho}(x) = xF_R(F_{\rho}(x))$$

- ▶ We'll find a similar result for the subset of nodes that are vulnerable.
- ▶ This is a node-based percolation problem.
- ▶ For a general monotonic threshold distribution  $f(\phi)$ , a degree  $k$  node is vulnerable with probability

$$B_{k\gamma} = \int_0^{1/k} f(\phi) d\phi.$$

Contagion

Basic Contagion Models

Global spreading condition

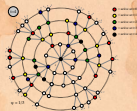
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **First goal:** Find the largest component of vulnerable nodes.
- ▶ Recall that for finding the giant component's size, we had to solve:

$$F_{\pi}(x) = xF_P(F_{\rho}(x)) \quad \text{and} \quad F_{\rho}(x) = xF_R(F_{\rho}(x))$$

- ▶ We'll find a similar result for the subset of nodes that are vulnerable.
- ▶ This is a node-based percolation problem.
- ▶ For a general monotonic threshold distribution  $f(\phi)$ , a degree  $k$  node is vulnerable with probability

$$B_{k\gamma} = \int_0^{1/k} f(\phi) d\phi.$$

Contagion

Basic Contagion Models

Global spreading condition

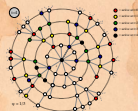
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **First goal:** Find the largest component of vulnerable nodes.
- ▶ Recall that for finding the giant component's size, we had to solve:

$$F_{\pi}(x) = xF_P(F_{\rho}(x)) \quad \text{and} \quad F_{\rho}(x) = xF_R(F_{\rho}(x))$$

- ▶ We'll find a similar result for the subset of nodes that are vulnerable.
- ▶ This is a node-based percolation problem.
- ▶ For a general monotonic threshold distribution  $f(\phi)$ , a degree  $k$  node is vulnerable with probability

$$B_{k\gamma} = \int_0^{1/k} f(\phi) d\phi.$$

Contagion

Basic Contagion Models

Global spreading condition

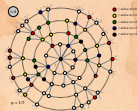
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **First goal:** Find the largest component of vulnerable nodes.
- ▶ Recall that for finding the giant component's size, we had to solve:

$$F_{\pi}(x) = xF_P(F_{\rho}(x)) \quad \text{and} \quad F_{\rho}(x) = xF_R(F_{\rho}(x))$$

- ▶ We'll find a similar result for the subset of nodes that are vulnerable.
- ▶ This is a node-based percolation problem.
- ▶ For a general monotonic threshold distribution  $f(\phi)$ , a degree  $k$  node is vulnerable with probability

$$B_{k1} = \int_0^{1/k} f(\phi) d\phi.$$

Contagion

Basic Contagion Models

Global spreading condition

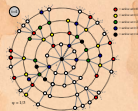
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Everything now revolves around the **modified** generating function:

$$F_P^{(\text{vuln})}(x) = \sum_{k=0}^{\infty} B_{k1} P_k x^k.$$

- ▶ Generating function for friends-of-friends distribution is related in same way as before:

$$F_A^{(\text{vuln})}(x) = \frac{\frac{d}{dx} F_P^{(\text{vuln})}(x)}{\frac{d}{dx} F_P^{(\text{vuln})}(x)|_{x=1}}$$

Contagion

Basic Contagion Models

Global spreading condition

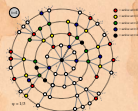
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Everything now revolves around the **modified** generating function:

$$F_P^{(\text{vuln})}(x) = \sum_{k=0}^{\infty} B_{k1} P_k x^k.$$

- ▶ Generating function for friends-of-friends distribution is related in same way as before:

$$F_R^{(\text{vuln})}(x) = \frac{\frac{d}{dx} F_P^{(\text{vuln})}(x)}{\frac{d}{dx} F_P^{(\text{vuln})}(x)|_{x=1}}.$$

Contagion

Basic Contagion Models

Global spreading condition

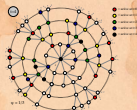
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\tau}^{(\text{vuln})}(x) = \underbrace{1 - F_P^{(\text{vuln})}(1)}_{\text{central node is not vulnerable}} + x F_P^{(\text{vuln})} \left( F_P^{(\text{vuln})}(x) \right)$$

central node  
is not  
vulnerable

$$F_P^{(\text{vuln})}(x) = \underbrace{1 - F_R^{(\text{vuln})}(1)}_{\text{first node is not vulnerable}} + x F_R^{(\text{vuln})} \left( F_P^{(\text{vuln})}(x) \right)$$

first node  
is not  
vulnerable

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\tau}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

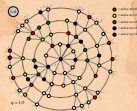
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\pi}^{(\text{vuln})}(x) = \underbrace{1 - F_{\rho}^{(\text{vuln})}(1)}_{\text{central node is not vulnerable}} + xF_{\rho}^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

central node  
is not  
vulnerable

$$F_{\rho}^{(\text{vuln})}(x) = \underbrace{1 - F_{\rho}^{(\text{vuln})}(1)}_{\text{first node is not vulnerable}} + xF_{\rho}^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

first node  
is not  
vulnerable

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\pi}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

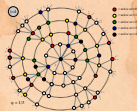
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\pi}^{(\text{vuln})}(x) = \underbrace{1 - F_P^{(\text{vuln})}(1)}_{\substack{\text{central node} \\ \text{is not} \\ \text{vulnerable}}} + xF_P^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

$$F_{\rho}^{(\text{vuln})}(x) = \underbrace{1 - F_R^{(\text{vuln})}(1)}_{\substack{\text{first node} \\ \text{is not} \\ \text{vulnerable}}} + xF_R^{(\text{vuln})}\left(F_{\pi}^{(\text{vuln})}(x)\right)$$

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\pi}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion Models

Global spreading condition

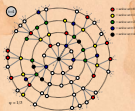
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\pi}^{(\text{vuln})}(x) = \underbrace{1 - F_P^{(\text{vuln})}(1)}_{\substack{\text{central node} \\ \text{is not} \\ \text{vulnerable}}} + xF_P^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

$$F_{\rho}^{(\text{vuln})}(x) = \underbrace{1 - F_R^{(\text{vuln})}(1)}_{\substack{\text{first node} \\ \text{is not} \\ \text{vulnerable}}} + xF_R^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\pi}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion  
Models

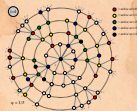
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\pi}^{(\text{vuln})}(x) = \underbrace{1 - F_P^{(\text{vuln})}(1)}_{\substack{\text{central node} \\ \text{is not} \\ \text{vulnerable}}} + xF_P^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

$$F_{\rho}^{(\text{vuln})}(x) = \underbrace{1 - F_R^{(\text{vuln})}(1)}_{\substack{\text{first node} \\ \text{is not} \\ \text{vulnerable}}} + xF_R^{(\text{vuln})}\left(F_{\rho}^{(\text{vuln})}(x)\right)$$

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\pi}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion  
Models

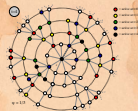
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ Functional relations for component size g.f.'s are almost the same...

$$F_{\pi}^{(\text{vuln})}(x) = \underbrace{1 - F_P^{(\text{vuln})}(1)}_{\substack{\text{central node} \\ \text{is not} \\ \text{vulnerable}}} + xF_P^{(\text{vuln})} \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

$$F_{\rho}^{(\text{vuln})}(x) = \underbrace{1 - F_R^{(\text{vuln})}(1)}_{\substack{\text{first node} \\ \text{is not} \\ \text{vulnerable}}} + xF_R^{(\text{vuln})} \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

- ▶ Can now solve as before to find  $S_{\text{vuln}} = 1 - F_{\pi}^{(\text{vuln})}(1)$ .

Contagion

Basic Contagion Models

Global spreading condition

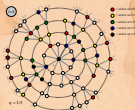
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Second goal:** Find probability of triggering largest vulnerable component.
- ▶ Assumption is first node is randomly chosen.
- ▶ Same set up as for vulnerable component except now we don't care if the initial node is vulnerable or not:

$$F_x^{(\text{big})}(x) = xF_P \left( F_P^{(\text{vuln})}(x) \right)$$

$$F_P^{(\text{vuln})}(x) = 1 - F_R'(1) + xF_R^{(\text{vuln})} \left( F_P^{(\text{vuln})}(x) \right)$$

- ▶ Solve as before to find  $P_{\text{big}} = S_{\text{big}} = 1 - F_x^{(\text{big})}(1)$ .

Contagion

Basic Contagion Models

Global spreading condition

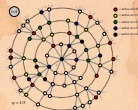
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Second goal:** Find probability of triggering largest vulnerable component.
- ▶ Assumption is **first node** is **randomly chosen**.
- ▶ Same set up as for vulnerable component except now we don't care if the initial node is vulnerable or not:

$$F_x^{(\text{big})}(x) = xF_P \left( F_P^{(\text{vuln})}(x) \right)$$

$$F_P^{(\text{vuln})}(x) = 1 - F_R'(1) + xF_R \left( F_P^{(\text{vuln})}(x) \right)$$

- ▶ Solve as before to find  $P_{\text{big}} = S_{\text{big}} = 1 - F_x^{(\text{big})}(1)$ .

Contagion

Basic Contagion Models

Global spreading condition

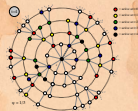
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Threshold contagion on random networks

- ▶ **Second goal:** Find probability of triggering largest vulnerable component.
- ▶ Assumption is **first node** is **randomly chosen**.
- ▶ **Same set up** as for vulnerable component except now we don't care if the initial node is vulnerable or not:

$$F_{\pi}^{(\text{trig})}(x) = x F_P \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

$$F_{\rho}^{(\text{vuln})}(x) = 1 - F_R^v(1) + x F_R^{(\text{vuln})} \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

- ▶ Solve as before to find  $P_{\text{big}} = S_{\text{big}} = 1 - F_{\pi}^{(\text{trig})}(1)$ .

Contagion

Basic Contagion Models

Global spreading condition

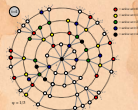
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Second goal:** Find probability of triggering largest vulnerable component.
- ▶ Assumption is **first node** is **randomly chosen**.
- ▶ **Same set up** as for vulnerable component except now we don't care if the initial node is vulnerable or not:

$$F_{\pi}^{(\text{trig})}(x) = x F_P \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

$$F_{\rho}^{(\text{vuln})}(x) = 1 - F_R^{\vee}(1) + x F_R^{(\text{vuln})} \left( F_{\rho}^{(\text{vuln})}(x) \right)$$

- ▶ Solve as before to find  $P_{\text{trig}} = S_{\text{trig}} = 1 - F_{\pi}^{(\text{trig})}(1)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

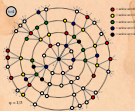
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Third goal:** Find expected fractional size of spread.
- ▶ Not obvious even for uniform threshold problem.
- ▶ Difficulty is in figuring out if and when nodes that need  $\geq 2$  hits switch on.
- ▶ Problem solved for infinite seed case by Gleeson and Cahalane:  
"Seed size strongly affects cascades on random networks," Phys. Rev. E, 2007. [7]
- ▶ Developed further by Gleeson in "Cascades on correlated and modular random networks," Phys. Rev. E, 2008. [6]

Contagion

Basic Contagion Models

Global spreading condition

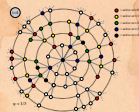
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Third goal:** Find expected fractional size of spread.
- ▶ Not obvious even for uniform threshold problem.
- ▶ Difficulty is in figuring out if and when nodes that need  $\geq 2$  hits switch on.
- ▶ Problem solved for infinite seed case by Gleeson and Cahalane:  
"Seed size strongly affects cascades on random networks," Phys. Rev. E, 2007. [7]
- ▶ Developed further by Gleeson in "Cascades on correlated and modular random networks," Phys. Rev. E, 2008. [6]

Contagion

Basic Contagion Models

Global spreading condition

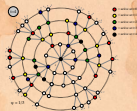
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Third goal:** Find expected fractional size of spread.
- ▶ Not obvious even for uniform threshold problem.
- ▶ Difficulty is in figuring out if and when nodes that need  $\geq 2$  hits switch on.
- ▶ Problem solved for infinite seed case by Gleeson and Cahalane:  
"Seed size strongly affects cascades on random networks," Phys. Rev. E, 2007. [7]
- ▶ Developed further by Gleeson in "Cascades on correlated and modular random networks," Phys. Rev. E, 2008. [6]

Contagion

Basic Contagion Models

Global spreading condition

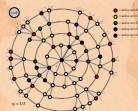
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Third goal:** Find expected fractional size of spread.
- ▶ Not obvious even for uniform threshold problem.
- ▶ Difficulty is in figuring out if and when nodes that need  $\geq 2$  hits switch on.
- ▶ Problem **solved** for infinite seed case by Gleeson and Cahalane:  
“Seed size strongly affects cascades on random networks,” Phys. Rev. E, 2007. [7]
- ▶ Developed further by Gleeson in “Cascades on correlated and modular random networks,” Phys. Rev. E, 2008. [6]

Contagion

Basic Contagion Models

Global spreading condition

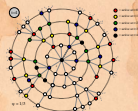
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Threshold contagion on random networks

- ▶ **Third goal:** Find expected fractional size of spread.
- ▶ Not obvious even for uniform threshold problem.
- ▶ Difficulty is in figuring out if and when nodes that need  $\geq 2$  hits switch on.
- ▶ Problem **solved** for infinite seed case by Gleeson and Cahalane:  
“Seed size strongly affects cascades on random networks,” Phys. Rev. E, 2007. [7]
- ▶ Developed further by Gleeson in “Cascades on correlated and modular random networks,” Phys. Rev. E, 2008. [6]

Contagion

Basic Contagion Models

Global spreading condition

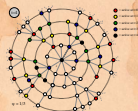
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

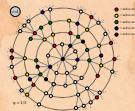
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

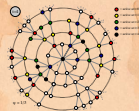
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

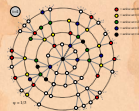
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

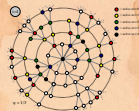
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

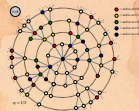
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

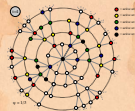
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Idea:

- ▶ Randomly turn on a fraction  $\phi_0$  of nodes at time  $t = 0$
- ▶ Capitalize on local branching network structure of random networks (again)
- ▶ Now think about what must happen for a specific node  $i$  to become active at time  $t$ :
  - $t = 0$ :  $i$  is one of the seeds (prob =  $\phi_0$ )
  - $t = 1$ :  $i$  was not a seed but enough of  $i$ 's friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = 2$ : enough of  $i$ 's friends and friends-of-friends switched on at time  $t = 0$  so that  $i$ 's threshold is now exceeded.
  - $t = n$ : enough nodes within  $n$  hops of  $i$  switched on at  $t = 0$  and their effects have propagated to reach  $i$ .

Contagion

Basic Contagion Models

Global spreading condition

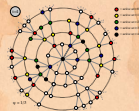
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

Contagion

Basic Contagion Models

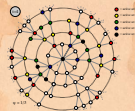
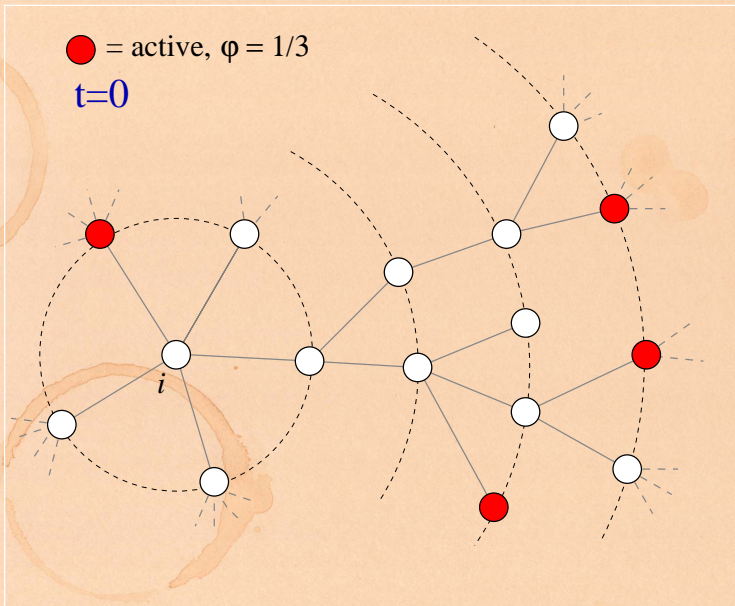
Global spreading condition

Social Contagion Models

Network version  
All-to-all networks

Theory

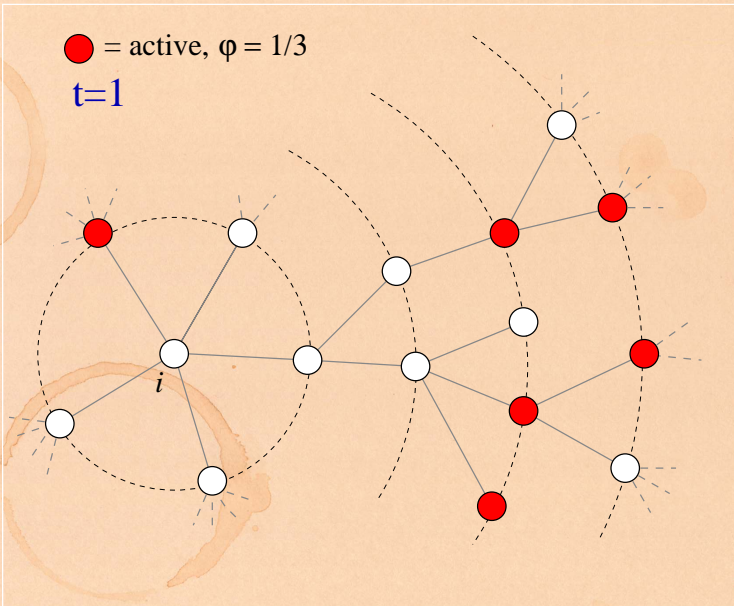
References



# Expected size of spread

● = active,  $\phi = 1/3$

$t=1$



Contagion

Basic Contagion Models

Global spreading condition

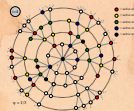
Social Contagion Models

Network version

All-to-all networks

Theory

References

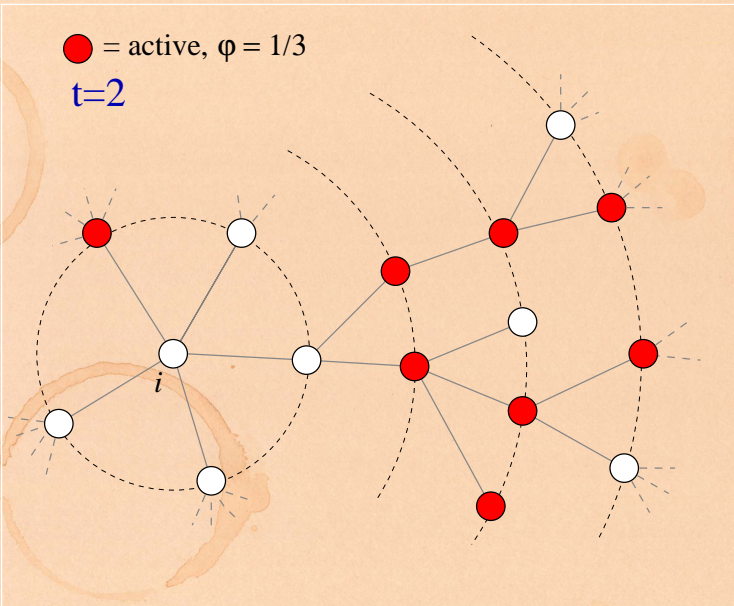




# Expected size of spread

● = active,  $\phi = 1/3$

$t=2$



Contagion

Basic Contagion  
Models

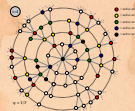
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# Expected size of spread

Contagion

Basic Contagion Models

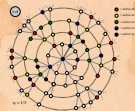
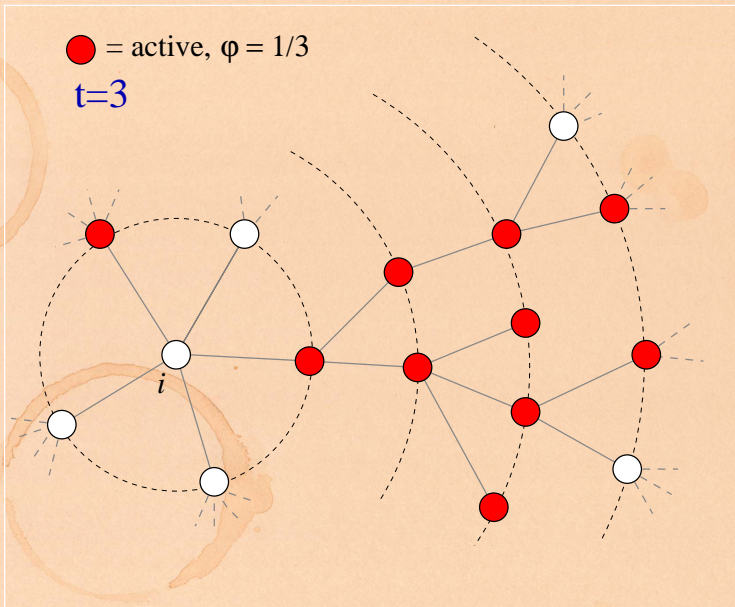
Global spreading condition

Social Contagion Models

Network version  
All-to-all networks

Theory

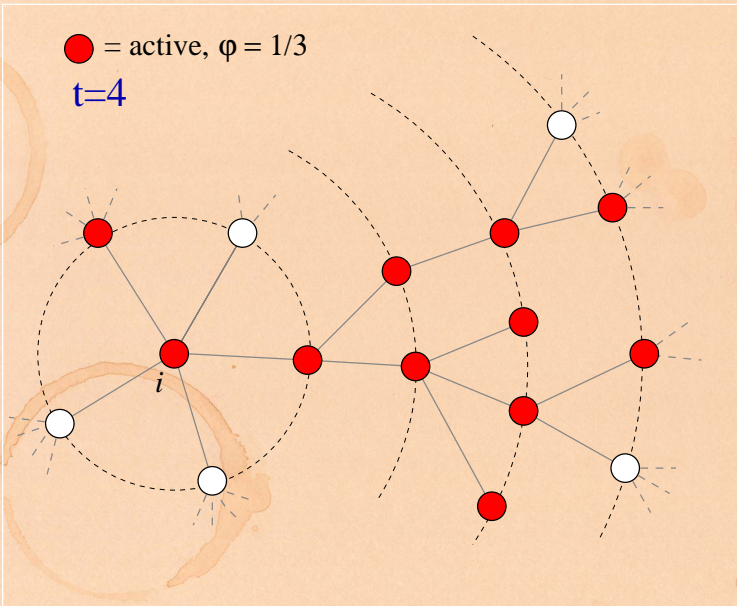
References



# Expected size of spread

● = active,  $\phi = 1/3$

$t=4$



Contagion

Basic Contagion Models

Global spreading condition

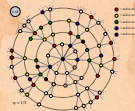
Social Contagion Models

Network version

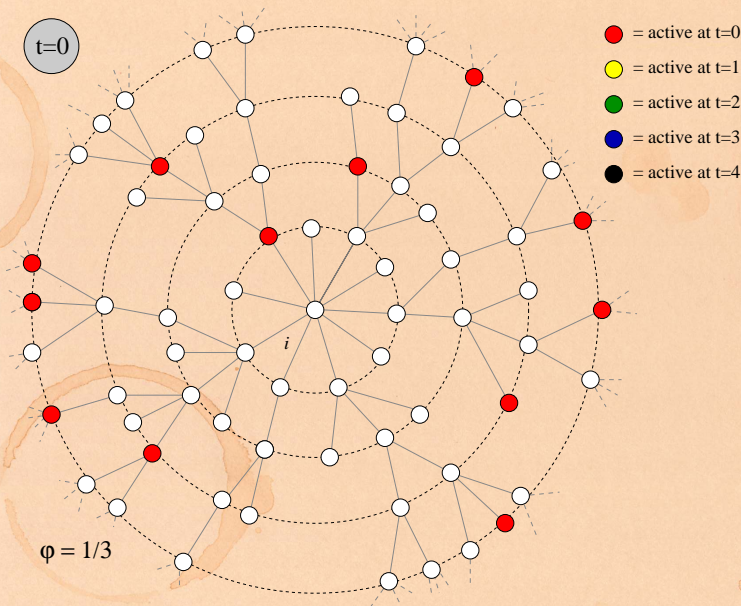
All-to-all networks

Theory

References



# Expected size of spread



Contagion

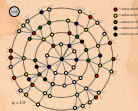
Basic Contagion Models

Global spreading condition

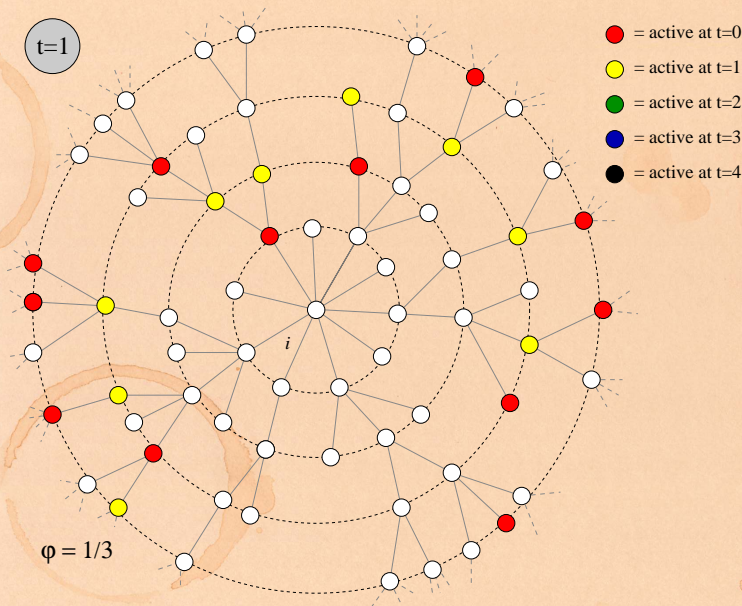
Social Contagion Models

Network version  
All-to-all networks  
Theory

References



# Expected size of spread



Contagion

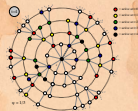
Basic Contagion  
Models

Global spreading  
condition

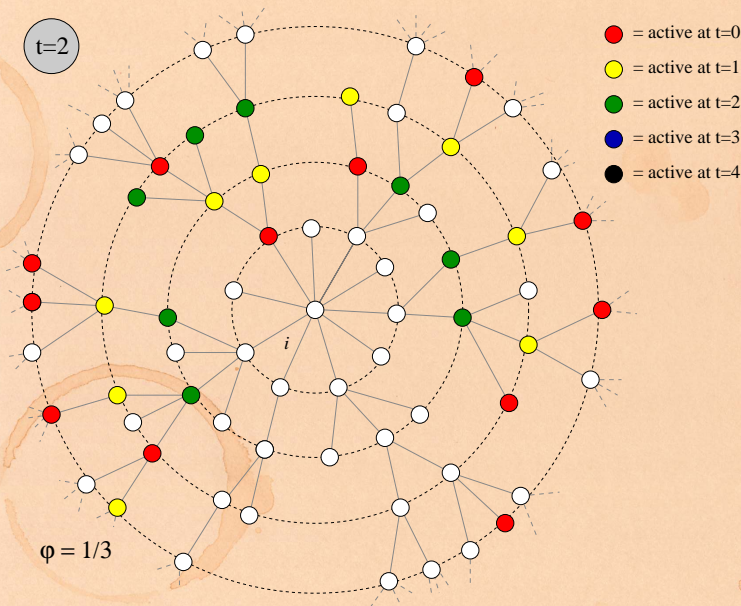
Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Expected size of spread



Contagion

Basic Contagion  
Models

Global spreading  
condition

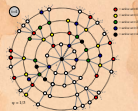
Social Contagion  
Models

Network version

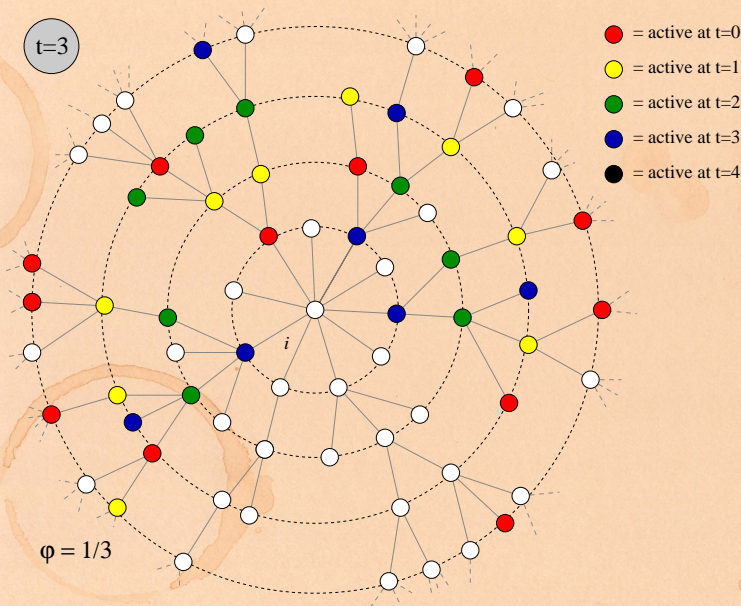
All-to-all networks

Theory

References



# Expected size of spread



Contagion

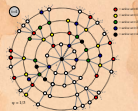
Basic Contagion  
Models

Global spreading  
condition

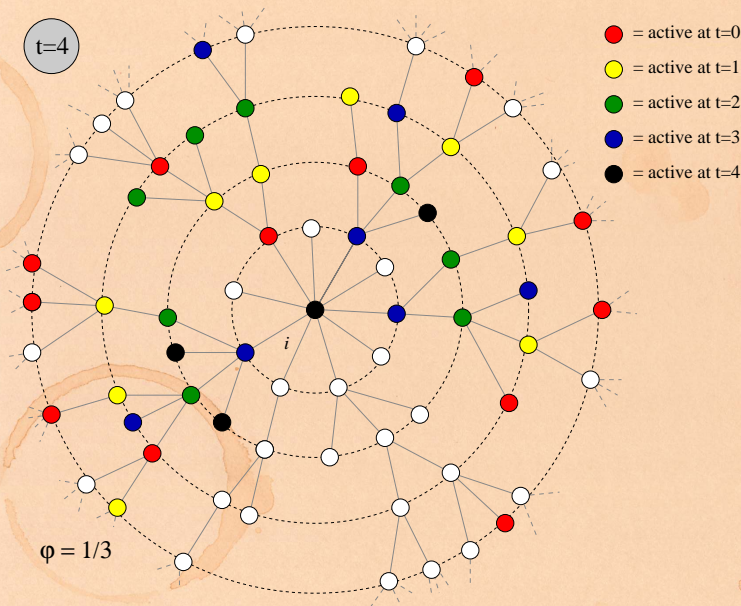
Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References



# Expected size of spread



Contagion

Basic Contagion Models

Global spreading condition

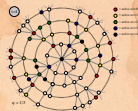
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Expected size of spread

## Notes:

- ▶ Calculations are possible nodes do not become inactive (strong restriction).
- ▶ Not just for threshold model—works for a wide range of contagion processes.
- ▶ We can analytically determine the entire time evolution, not just the final size.
- ▶ We can in fact determine  $\Pr(\text{node of degree } k \text{ switching on at time } t)$ .
- ▶ Asynchronous updating can be handled too.

Contagion

Basic Contagion Models

Global spreading condition

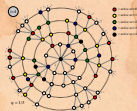
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Notes:

- ▶ Calculations are possible nodes do not become inactive (strong restriction).
- ▶ Not just for threshold model—works for a wide range of contagion processes.
- ▶ We can analytically determine the entire time evolution, not just the final size.
- ▶ We can in fact determine  $\Pr(\text{node of degree } k \text{ switching on at time } t)$ .
- ▶ Asynchronous updating can be handled too.

Contagion

Basic Contagion Models

Global spreading condition

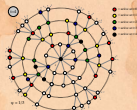
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Notes:

- ▶ Calculations are possible nodes do not become inactive (strong restriction).
- ▶ Not just for threshold model—works for a wide range of contagion processes.
- ▶ We can analytically determine the entire time evolution, not just the final size.
- ▶ We can in fact determine  $\Pr(\text{node of degree } k \text{ switching on at time } t)$ .
- ▶ Asynchronous updating can be handled too.

Contagion

Basic Contagion Models

Global spreading condition

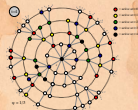
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Notes:

- ▶ Calculations are possible nodes do not become inactive (strong restriction).
- ▶ Not just for threshold model—works for a wide range of contagion processes.
- ▶ We can analytically determine the entire time evolution, not just the final size.
- ▶ We can in fact determine  $\Pr(\text{node of degree } k \text{ switching on at time } t)$ .
- ▶ Asynchronous updating can be handled too.

Contagion

Basic Contagion Models

Global spreading condition

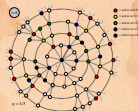
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Notes:

- ▶ Calculations are possible nodes do not become inactive (strong restriction).
- ▶ Not just for threshold model—works for a wide range of contagion processes.
- ▶ We can analytically determine the entire time evolution, not just the final size.
- ▶ We can in fact determine  $\Pr(\text{node of degree } k \text{ switching on at time } t)$ .
- ▶ Asynchronous updating can be handled too.

Contagion

Basic Contagion Models

Global spreading condition

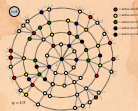
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Pleasantness:

- ▶ Taking off from a single seed story is about **expansion** away from a node.
- ▶ Extent of spreading story is about contraction at a node.

Contagion

Basic Contagion Models

Global spreading condition

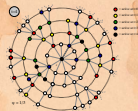
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## Pleasantness:

- ▶ Taking off from a single seed story is about **expansion** away from a node.
- ▶ Extent of spreading story is about **contraction** at a node.

Contagion

Basic Contagion  
Models

Global spreading  
condition

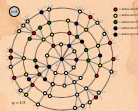
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ► Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

► Notation:  $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

► Our starting point:  $\phi_{k,0} = \phi_0.$

►  $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

► Probability a degree  $k$  node was a seed at  $t = 0$  is  $\phi_0$  (as above).

► Probability a degree  $k$  node was not a seed at  $t = 0$  is  $(1 - \phi_0).$

► Combining everything, we have:

$$\phi_{k,1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

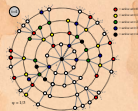
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Expected size of spread

## ▶ Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

## ▶ Notation: $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

▶ Our starting point:  $\phi_{k,0} = \phi_0.$

▶  $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

▶ Probability a degree  $k$  node was a seed at  $t = 0$  is  $\phi_0$  (as above).

▶ Probability a degree  $k$  node was not a seed at  $t = 0$  is  $(1 - \phi_0).$

▶ Combining everything, we have:

$$\phi_{k,t} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

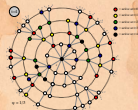
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ▶ Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

## ▶ Notation: $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

## ▶ Our starting point: $\phi_{k,0} = \phi_0.$

▶  $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

▶ Probability a degree  $k$  node was a seed at  $t = 0$  is  $\phi_0$  (as above).

▶ Probability a degree  $k$  node was not a seed at  $t = 0$  is  $(1 - \phi_0).$

▶ Combining everything, we have:

$$\phi_{k,t} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

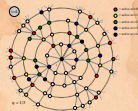
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ▶ Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

## ▶ Notation: $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

## ▶ Our starting point: $\phi_{k,0} = \phi_0.$

## ▶ $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

## ▶ Probability a degree $k$ node was a seed at $t = 0$ is $\phi_0$ (as above).

## ▶ Probability a degree $k$ node was not a seed at $t = 0$ is $(1 - \phi_0).$

## ▶ Combining everything, we have:

$$\phi_{k,t} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

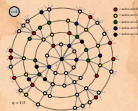
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ▶ Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

## ▶ Notation: $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

## ▶ Our starting point: $\phi_{k,0} = \phi_0.$

## ▶ $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

## ▶ Probability a degree $k$ node was a seed at $t = 0$ is $\phi_0$ (as above).

## ▶ Probability a degree $k$ node was not a seed at $t = 0$ is $(1 - \phi_0).$

## ▶ Combining everything, we have:

$$\phi_{k,t} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

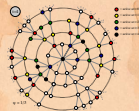
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ▶ Notation:

$\phi_{k,t} = \Pr(\text{a degree } k \text{ node is active at time } t).$

## ▶ Notation: $B_{kj} = \Pr(\text{a degree } k \text{ node becomes active if } j \text{ neighbors are active}).$

## ▶ Our starting point: $\phi_{k,0} = \phi_0.$

## ▶ $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \Pr(j \text{ of a degree } k \text{ node's neighbors were seeded at time } t = 0).$

## ▶ Probability a degree $k$ node was a seed at $t = 0$ is $\phi_0$ (as above).

## ▶ Probability a degree $k$ node was not a seed at $t = 0$ is $(1 - \phi_0).$

## ▶ Combining everything, we have:

$$\phi_{k,t} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

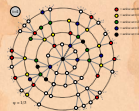
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

## ► Notation:

$\phi_{k,t} = \mathbf{Pr}$ (a degree  $k$  node is active at time  $t$ ).

## ► Notation: $B_{kj} = \mathbf{Pr}$ (a degree $k$ node becomes active if $j$ neighbors are active).

## ► Our starting point: $\phi_{k,0} = \phi_0$ .

## ► $\binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} = \mathbf{Pr}$ ( $j$ of a degree $k$ node's neighbors were seeded at time $t = 0$ ).

## ► Probability a degree $k$ node was a seed at $t = 0$ is $\phi_0$ (as above).

## ► Probability a degree $k$ node was not a seed at $t = 0$ is $(1 - \phi_0)$ .

## ► Combining everything, we have:

$$\phi_{k,1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^k \binom{k}{j} \phi_0^j (1 - \phi_0)^{k-j} B_{kj}.$$

Contagion

Basic Contagion Models

Global spreading condition

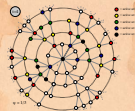
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i - j} B_{k_i, j}$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t,t+1}$ :

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k, j}$$

- ▶ So we need to compute  $\theta_t, \dots$

Contagion

Basic Contagion  
Models

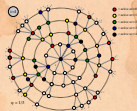
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_j} \binom{k_j}{j} \theta_t^j (1 - \theta_t)^{k_j - j} B_{k_j, j}$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t,t+1}$ :

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k, j}$$

- ▶ So we need to compute  $\theta_t, \dots$

Contagion

Basic Contagion  
Models

Global spreading  
condition

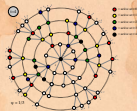
Social Contagion  
Models

Network version

All-to-all networks

Theory

References





# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i - j} B_{k_i, j}$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t,t+1}$ :

$$\phi_{t,t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k, j}$$

- ▶ So we need to compute  $\theta_t, \dots$

Contagion

Basic Contagion Models

Global spreading condition

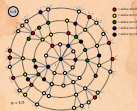
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{i,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i-j} B_{k_i j}.$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t+1}$ :

$$\phi_{t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k j}.$$

- ▶ So we need to compute  $\theta_t, \dots$

Contagion

Basic Contagion Models

Global spreading condition

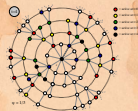
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{i,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i-j} B_{k_i j}.$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t+1}$ :

$$\phi_{t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k j}.$$

- ▶ So we need to compute  $\theta_t, \dots$

Contagion

Basic Contagion  
Models

Global spreading  
condition

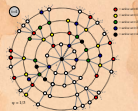
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{i,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i-j} B_{k_i j}.$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t+1}$ :

$$\phi_{t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k j}.$$

- ▶ So we need to compute  $\theta_t$ ...

Contagion

Basic Contagion  
Models

Global spreading  
condition

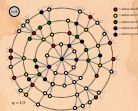
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

- ▶ For general  $t$ , we need to know the probability an edge coming into a degree  $k$  node at time  $t$  is active.
- ▶ **Notation:** call this probability  $\theta_t$ .
- ▶ We already know  $\theta_0 = \phi_0$ .
- ▶ Story analogous to  $t = 1$  case:

$$\phi_{i,t+1} = \phi_0 + (1 - \phi_0) \sum_{j=0}^{k_i} \binom{k_i}{j} \theta_t^j (1 - \theta_t)^{k_i-j} B_{k_i j}.$$

- ▶ Average over all nodes to obtain expression for  $\phi_{t+1}$ :

$$\phi_{t+1} = \phi_0 + (1 - \phi_0) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{k j}.$$

- ▶ So we need to compute  $\theta_t$ ... massive excitement...

Contagion

Basic Contagion  
Models

Global spreading  
condition

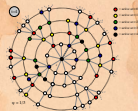
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Expected size of spread

Contagion

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

All-to-all networks

Theory

References

First connect  $\theta_0$  to  $\theta_1$ :

▶  $\theta_1 = \phi_0 +$

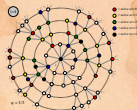
$$(1 - \phi_0) \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \sum_{j=0}^{k-1} \binom{k-1}{j} \theta_0^j (1 - \theta_0)^{k-1-j} B_{kj}$$

▶  $\frac{kP_k}{\langle k \rangle} = R_k = \mathbf{Pr}$  (edge connects to a degree  $k$  node).

▶  $\sum_{j=0}^{k-1}$  piece gives  $\mathbf{Pr}$ (degree node  $k$  activates) of its neighbors  $k - 1$  incoming neighbors are active.

▶  $\phi_0$  and  $(1 - \phi_0)$  terms account for state of node at time  $t = 0$ .

▶ See this all generalizes to give  $\theta_{t+1}$  in terms of  $\theta_t, \dots$



# Expected size of spread

Contagion

Basic Contagion Models

Global spreading condition

Social Contagion Models

Network version

All-to-all networks

Theory

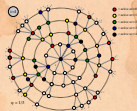
References

First connect  $\theta_0$  to  $\theta_1$ :

▶  $\theta_1 = \phi_0 +$

$$(1 - \phi_0) \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \sum_{j=0}^{k-1} \binom{k-1}{j} \theta_0^j (1 - \theta_0)^{k-1-j} B_{kj}$$

- ▶  $\frac{kP_k}{\langle k \rangle} = R_k = \mathbf{Pr}$  (edge connects to a degree  $k$  node).
- ▶  $\sum_{j=0}^{k-1}$  piece gives  $\mathbf{Pr}$ (degree node  $k$  activates) of its neighbors  $k - 1$  incoming neighbors are active.
- ▶  $\phi_0$  and  $(1 - \phi_0)$  terms account for state of node at time  $t = 0$ .
- ▶ See this all generalizes to give  $\theta_{t+1}$  in terms of  $\theta_t...$



# Expected size of spread

Two pieces: edges first, and then nodes

$$1. \theta_{t+1} = \underbrace{\phi_0}_{\text{exogenous}}$$

$$+(1 - \phi_0) \underbrace{\sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \sum_{j=0}^{k-1} \binom{k-1}{j} \theta_t^j (1 - \theta_t)^{k-1-j} B_{kj}}_{\text{social effects}}$$

with  $\theta_0 = \phi_0$ .

$$2. \phi_{t+1} =$$

$$\underbrace{\phi_0}_{\text{exogenous}} + (1 - \phi_0) \underbrace{\sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{kj}}_{\text{social effects}}$$

Contagion

Basic Contagion Models

Global spreading condition

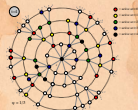
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Comparison between theory and simulations

Contagion

Basic Contagion  
Models

Global spreading  
condition

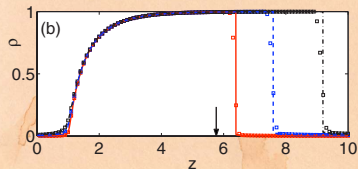
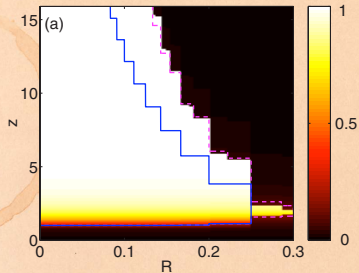
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



▶ Pure random networks with simple threshold responses

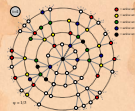
▶  $R =$  uniform threshold (our  $\phi_*$ );  $z =$  average degree;  $\rho = \phi$ ;  $q = \theta$ ;  $N = 10^5$ .

▶  $\phi_0 = 10^{-3}$ ,  $0.5 \times 10^{-2}$ , and  $10^{-2}$ .

▶ Cascade window is for  $\phi_0 = 10^{-2}$  case.

▶ Sensible expansion of cascade window as  $\phi_0$  increases.

From Gleeson and Cahalane [7]



# Notes:

- ▶ Retrieve cascade condition for spreading from a single seed in limit  $\phi_0 \rightarrow 0$ .
- ▶ Depends on map  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ First: if self-starters are present, some activation is assured:

$$G(0; \phi_0) = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet B_{k0} > 0.$$

meaning  $B_{k0} > 0$  for at least one value of  $k \geq 1$ .

- ▶ If  $\theta = 0$  is a fixed point of  $G$  (i.e.,  $G(0; \phi_0) = 0$ ) then spreading occurs if

$$G'(0; \phi_0) = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) \bullet B_{k1} > 1.$$

Insert question from assignment 8 (田)

Contagion

Basic Contagion Models

Global spreading condition

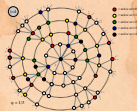
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

- ▶ Retrieve cascade condition for spreading from a single seed in limit  $\phi_0 \rightarrow 0$ .
- ▶ Depends on map  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ First: if self-starters are present, some activation is assured:

$$G(0; \phi_0) = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet B_{k0} > 0.$$

meaning  $B_{k0} > 0$  for at least one value of  $k \geq 1$ .

- ▶ If  $\theta = 0$  is a fixed point of  $G$  (i.e.,  $G(0; \phi_0) = 0$ ) then spreading occurs if

$$G'(0; \phi_0) = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) \bullet B_{k1} > 1.$$

Insert question from assignment 8 (田)

Contagion

Basic Contagion Models

Global spreading condition

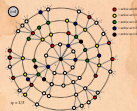
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

- ▶ Retrieve cascade condition for spreading from a single seed in limit  $\phi_0 \rightarrow 0$ .
- ▶ Depends on map  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ First: if self-starters are present, some activation is assured:

$$G(0; \phi_0) = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet B_{k0} > 0.$$

meaning  $B_{k0} > 0$  for at least one value of  $k \geq 1$ .

- ▶ If  $\theta = 0$  is a fixed point of  $G$  (i.e.,  $G(0; \phi_0) = 0$ ) then spreading occurs if

$$G'(0; \phi_0) = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k-1) \bullet B_{k1} > 1.$$

Insert question from assignment 8 (田)

Contagion

Basic Contagion Models

Global spreading condition

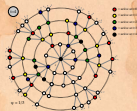
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

- ▶ Retrieve cascade condition for spreading from a single seed in limit  $\phi_0 \rightarrow 0$ .
- ▶ Depends on map  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ First: if self-starters are present, some activation is assured:

$$G(0; \phi_0) = \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet B_{k0} > 0.$$

meaning  $B_{k0} > 0$  for at least one value of  $k \geq 1$ .

- ▶ If  $\theta = 0$  is a fixed point of  $G$  (i.e.,  $G(0; \phi_0) = 0$ ) then spreading occurs if

$$G'(0; \phi_0) = \sum_{k=0}^{\infty} \frac{kP_k}{\langle k \rangle} \bullet (k - 1) \bullet B_{k1} > 1.$$

Insert question from assignment 8 (田)

Contagion

Basic Contagion Models

Global spreading condition

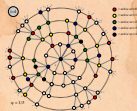
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

## In words:

- ▶ If  $G(0; \phi_0) > 0$ , spreading must occur because some nodes turn on for free.
- ▶ If  $G$  has an unstable fixed point at  $\theta = 0$ , then cascades are also always possible.

## Non-vanishing seed case:

- ▶ Cascade condition is more complicated for  $\phi_0 > 0$ .
- ▶ If  $G$  has a stable fixed point at  $\theta = 0$ , and an unstable fixed point for some  $0 < \theta_* < 1$ , then for  $\theta_0 > \theta_*$ , spreading takes off.
- ▶ Tricky point:  $G$  depends on  $\phi_0$ , so as we change  $\phi_0$ , we also change  $G$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

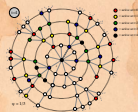
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Notes:

## In words:

- ▶ If  $G(0; \phi_0) > 0$ , spreading must occur because some nodes turn on for free.
- ▶ If  $G$  has an **unstable fixed point** at  $\theta = 0$ , then cascades are also always possible.

## Non-vanishing seed case:

- ▶ Cascade condition is more complicated for  $\phi_0 > 0$ .
- ▶ If  $G$  has a stable fixed point at  $\theta = 0$ , and an unstable fixed point for some  $0 < \theta_* < 1$ , then for  $\theta_0 > \theta_*$ , spreading takes off.
- ▶ Tricky point:  $G$  depends on  $\phi_0$ , so as we change  $\phi_0$ , we also change  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

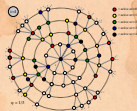
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

## In words:

- ▶ If  $G(0; \phi_0) > 0$ , spreading must occur because some nodes turn on for free.
- ▶ If  $G$  has an **unstable fixed point** at  $\theta = 0$ , then cascades are also always possible.

## Non-vanishing seed case:

- ▶ Cascade condition is more complicated for  $\phi_0 > 0$ .
- ▶ If  $G$  has a stable fixed point at  $\theta = 0$ , and an unstable fixed point for some  $0 < \theta_* < 1$ , then for  $\theta_0 > \theta_*$ , spreading takes off.
- ▶ Tricky point:  $G$  depends on  $\phi_0$ , so as we change  $\phi_0$ , we also change  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

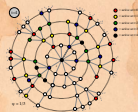
Social Contagion Models

Network version

All-to-all networks

Theory

References





# Notes:

## In words:

- ▶ If  $G(0; \phi_0) > 0$ , spreading must occur because some nodes turn on for free.
- ▶ If  $G$  has an **unstable fixed point** at  $\theta = 0$ , then cascades are also always possible.

## Non-vanishing seed case:

- ▶ Cascade condition is more complicated for  $\phi_0 > 0$ .
- ▶ If  $G$  has a **stable fixed point** at  $\theta = 0$ , and an **unstable fixed point** for some  $0 < \theta_* < 1$ , then for  $\theta_0 > \theta_*$ , spreading takes off.
- ▶ Tricky point:  $G$  depends on  $\phi_0$ , so as we change  $\phi_0$ , we also change  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

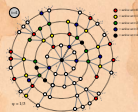
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Notes:

## In words:

- ▶ If  $G(0; \phi_0) > 0$ , spreading must occur because some nodes turn on for free.
- ▶ If  $G$  has an **unstable fixed point** at  $\theta = 0$ , then cascades are also always possible.

## Non-vanishing seed case:

- ▶ Cascade condition is more complicated for  $\phi_0 > 0$ .
- ▶ If  $G$  has a **stable fixed point** at  $\theta = 0$ , and an **unstable fixed point** for some  $0 < \theta_* < 1$ , then for  $\theta_0 > \theta_*$ , spreading takes off.
- ▶ Tricky point:  $G$  depends on  $\phi_0$ , so as we change  $\phi_0$ , we also change  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

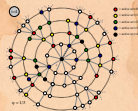
Social Contagion Models

Network version

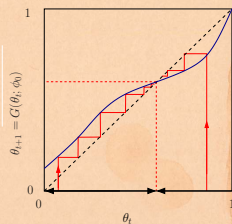
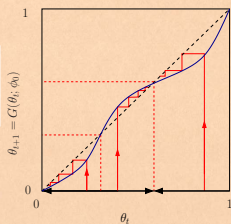
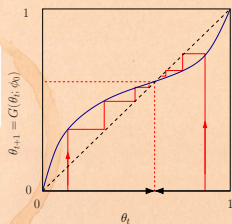
All-to-all networks

Theory

References



# General fixed point story:



- ▶ Given  $\theta_0 (= \phi_0)$ ,  $\theta_\infty$  will be the nearest stable fixed point, either above or below.
- ▶ n.b., adjacent fixed points must have opposite stability types.
- ▶ Important: Actual form of  $G$  depends on  $\phi_0$ .
- ▶ So choice of  $\phi_0$  dictates both  $G$  and starting point—can't start anywhere for a given  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

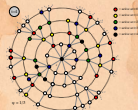
Social Contagion Models

Network version

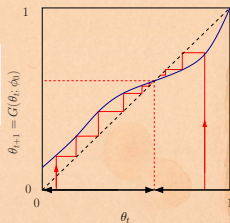
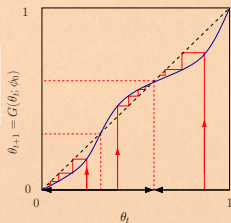
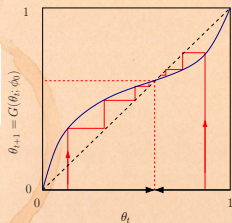
All-to-all networks

Theory

References



# General fixed point story:



- ▶ Given  $\theta_0 (= \phi_0)$ ,  $\theta_\infty$  will be the nearest stable fixed point, either above or below.
- ▶ n.b., adjacent fixed points must have opposite stability types.
- ▶ Important: Actual form of  $G$  depends on  $\phi_0$ .
- ▶ So choice of  $\phi_0$  dictates both  $G$  and starting point—can't start anywhere for a given  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

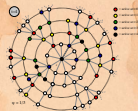
Social Contagion Models

Network version

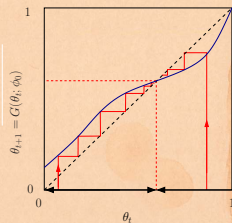
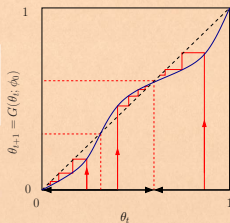
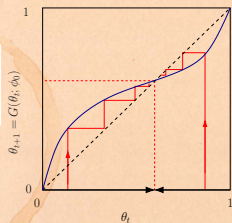
All-to-all networks

Theory

References



# General fixed point story:



- ▶ Given  $\theta_0 (= \phi_0)$ ,  $\theta_\infty$  will be the nearest stable fixed point, either above or below.
- ▶ n.b., adjacent fixed points must have opposite stability types.
- ▶ **Important:** Actual form of  $G$  depends on  $\phi_0$ .
- ▶ So choice of  $\phi_0$  dictates both  $G$  and starting point—can't start anywhere for a given  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

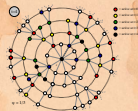
Social Contagion Models

Network version

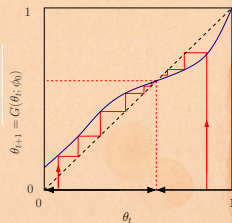
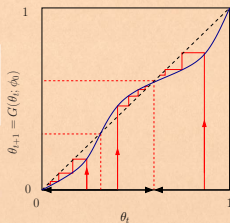
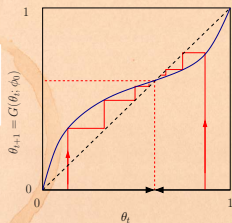
All-to-all networks

Theory

References



# General fixed point story:



- ▶ Given  $\theta_0 (= \phi_0)$ ,  $\theta_\infty$  will be the nearest stable fixed point, either above or below.
- ▶ n.b., adjacent fixed points must have opposite stability types.
- ▶ **Important:** Actual form of  $G$  depends on  $\phi_0$ .
- ▶ So choice of  $\phi_0$  dictates both  $G$  and starting point—can't start anywhere for a given  $G$ .

Contagion

Basic Contagion Models

Global spreading condition

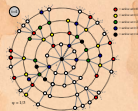
Social Contagion Models

Network version

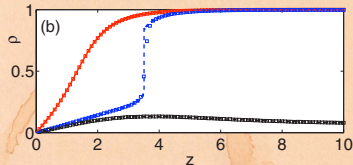
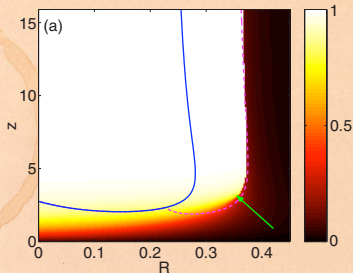
All-to-all networks

Theory

References



# Comparison between theory and simulations



- ▶ Now allow thresholds to be distributed according to a Gaussian with mean  $R$ .
- ▶  $R = 0.2$ ,  $0.362$ , and  $0.38$ ;  $\sigma = 0.2$ .
- ▶  $\phi_0 = 0$  but some nodes have thresholds  $\leq 0$  so effectively  $\phi_0 > 0$ .
- ▶ Now see a (nasty) discontinuous phase transition for low  $k$ .

Contagion

Basic Contagion Models

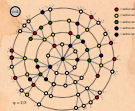
Global spreading condition

Social Contagion Models

Network version  
All-to-all networks

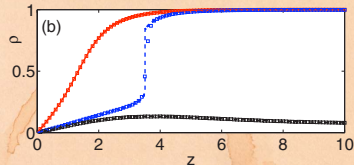
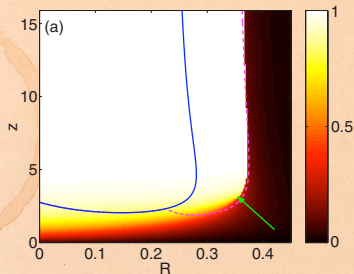
Theory

References



From Gleeson and Cahalane [7]

# Comparison between theory and simulations



From Gleeson and Cahalane [7]

- ▶ Now allow thresholds to be distributed according to a Gaussian with mean  $R$ .
- ▶  $R = 0.2$ ,  $0.362$ , and  $0.38$ ;  $\sigma = 0.2$ .
- ▶  $\phi_0 = 0$  but some nodes have thresholds  $\leq 0$  so effectively  $\phi_0 > 0$ .
- ▶ Now see a (nasty) discontinuous phase transition for low  $\langle k \rangle$ .

Contagion

Basic Contagion Models

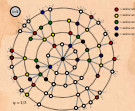
Global spreading condition

Social Contagion Models

Network version  
All-to-all networks

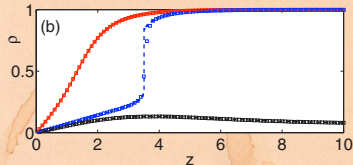
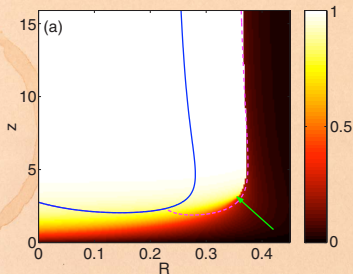
Theory

References





# Comparison between theory and simulations



From Gleeson and Cahalane [7]

- ▶ Now allow thresholds to be distributed according to a Gaussian with mean  $R$ .
- ▶  $R = 0.2$ ,  $0.362$ , and  $0.38$ ;  $\sigma = 0.2$ .
- ▶  $\phi_0 = 0$  but some nodes have thresholds  $\leq 0$  so effectively  $\phi_0 > 0$ .
- ▶ Now see a (nasty) discontinuous phase transition for low  $\langle k \rangle$ .

Contagion

Basic Contagion Models

Global spreading condition

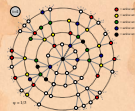
Social Contagion Models

Network version

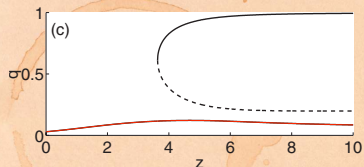
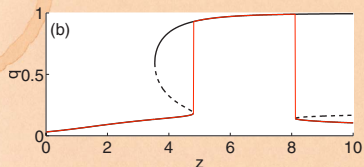
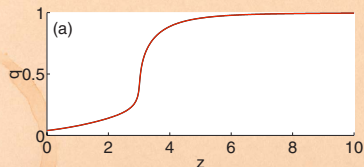
All-to-all networks

Theory

References



# Comparison between theory and simulations



- ▶ Plots of stability points for  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ n.b.: 0 is not a fixed point here:  $\theta_0 = 0$  always takes off.
- ▶ Top to bottom:  $R = 0.35, 0.371, \text{ and } 0.375$ .
- ▶ n.b.: higher values of  $\theta_0$  for (b) and (c) lead to higher fixed points of  $G$ .
- ▶ Saddle node bifurcations appear and merge (b and c).

Contagion

Basic Contagion Models

Global spreading condition

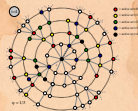
Social Contagion Models

Network version

All-to-all networks

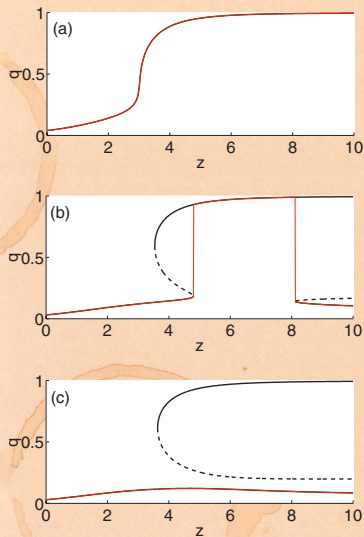
Theory

References



From Gleeson and Cahalane [7]

# Comparison between theory and simulations



- ▶ Plots of stability points for  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ n.b.: 0 is not a fixed point here:  $\theta_0 = 0$  always takes off.
- ▶ Top to bottom:  $R = 0.35, 0.371, \text{ and } 0.375$ .
- ▶ n.b.: higher values of  $\theta_0$  for (b) and (c) lead to higher fixed points of  $G$ .
- ▶ Saddle node bifurcations appear and merge (b and c).

Contagion

Basic Contagion Models

Global spreading condition

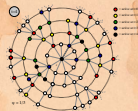
Social Contagion Models

Network version

All-to-all networks

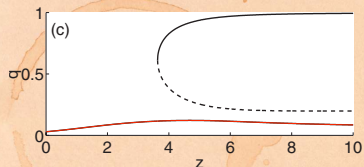
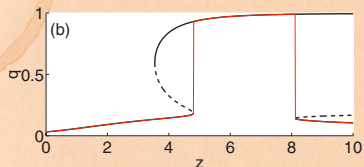
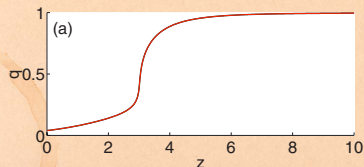
Theory

References



From Gleeson and Cahalane [7]

# Comparison between theory and simulations



- ▶ Plots of stability points for  $\theta_{t+1} = G(\theta_t; \phi_0)$ .
- ▶ n.b.: 0 is not a fixed point here:  $\theta_0 = 0$  always takes off.
- ▶ Top to bottom:  $R = 0.35, 0.371, \text{ and } 0.375$ .
- ▶ n.b.: higher values of  $\theta_0$  for (b) and (c) lead to higher fixed points of  $G$ .
- ▶ Saddle node bifurcations appear and merge (b and c).

Contagion

Basic Contagion Models

Global spreading condition

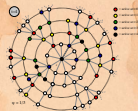
Social Contagion Models

Network version

All-to-all networks

Theory

References



From Gleeson and Cahalane [7]

# Spreadarama

## Bridging to single seed case:

- ▶ Consider largest vulnerable component as initial set of seeds.
- ▶ Not quite right as spreading must move through vulnerables.
- ▶ But we can usefully think of the vulnerable component as activating at time  $t = 0$  because order doesn't matter.
- ▶ Rebuild  $\phi_t$  and  $\theta_t$  expressions...

Contagion

Basic Contagion Models

Global spreading condition

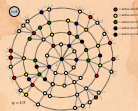
Social Contagion Models

Network version

All-to-all networks

Theory

References



# Spreadarama

## Bridging to single seed case:

- ▶ Consider largest vulnerable component as initial set of seeds.
- ▶ Not quite right as spreading must move through vulnerables.
- ▶ But we can usefully think of the vulnerable component as activating at time  $t = 0$  because order doesn't matter.
- ▶ Rebuild  $\phi_t$  and  $\theta_t$  expressions...

Contagion

Basic Contagion Models

Global spreading condition

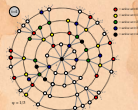
Social Contagion Models

Network version

All-to-all networks

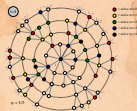
Theory

References



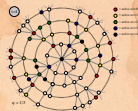
## Bridging to single seed case:

- ▶ Consider largest vulnerable component as initial set of seeds.
- ▶ Not quite right as spreading must move through vulnerables.
- ▶ But we can usefully think of the vulnerable component as activating at time  $t = 0$  because order doesn't matter.
- ▶ Rebuild  $\phi_t$  and  $\theta_t$  expressions...



## Bridging to single seed case:

- ▶ Consider largest vulnerable component as initial set of seeds.
- ▶ Not quite right as spreading must move through vulnerables.
- ▶ But we can usefully think of the vulnerable component as activating at time  $t = 0$  because order doesn't matter.
- ▶ Rebuild  $\phi_t$  and  $\theta_t$  expressions...





## Two pieces modified for single seed:

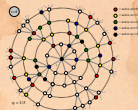
1.  $\theta_{t+1} = \theta_{\text{vuln}} +$

$$(1 - \theta_{\text{vuln}}) \sum_{k=1}^{\infty} \frac{kP_k}{\langle k \rangle} \sum_{j=0}^{k-1} \binom{k-1}{j} \theta_t^j (1 - \theta_t)^{k-1-j} B_{kj}$$

with  $\theta_0 = \theta_{\text{vuln}} = \mathbf{Pr}$  an edge leads to the giant vulnerable component (if it exists).

2.  $\phi_{t+1} = S_{\text{vuln}} +$

$$(1 - S_{\text{vuln}}) \sum_{k=0}^{\infty} P_k \sum_{j=0}^k \binom{k}{j} \theta_t^j (1 - \theta_t)^{k-j} B_{kj}.$$



# Time-dependent solutions

## Synchronous update

- ▶ Done: Evolution of  $\phi_t$  and  $\theta_t$  given exactly by the maps we have derived.

## Asynchronous updates

- ▶ Update nodes with probability  $\alpha$ .
- ▶ As  $\alpha \rightarrow 0$ , updates become effectively independent.
- ▶ Now can talk about  $\phi(t)$  and  $\theta(t)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

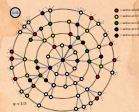
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Time-dependent solutions

## Synchronous update

- ▶ Done: Evolution of  $\phi_t$  and  $\theta_t$  given exactly by the maps we have derived.

## Asynchronous updates

- ▶ Update nodes with probability  $\alpha$ .
- ▶ As  $\alpha \rightarrow 0$ , updates become effectively independent.
- ▶ Now can talk about  $\phi(t)$  and  $\theta(t)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

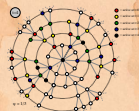
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Time-dependent solutions

## Synchronous update

- ▶ Done: Evolution of  $\phi_t$  and  $\theta_t$  given exactly by the maps we have derived.

## Asynchronous updates

- ▶ Update nodes with probability  $\alpha$ .
- ▶ As  $\alpha \rightarrow 0$ , updates become effectively independent.
- ▶ Now can talk about  $\phi(t)$  and  $\theta(t)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

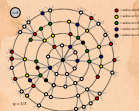
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Time-dependent solutions

## Synchronous update

- ▶ Done: Evolution of  $\phi_t$  and  $\theta_t$  given exactly by the maps we have derived.

## Asynchronous updates

- ▶ Update nodes with probability  $\alpha$ .
- ▶ As  $\alpha \rightarrow 0$ , updates become effectively independent.
- ▶ Now can talk about  $\phi(t)$  and  $\theta(t)$ .

Contagion

Basic Contagion  
Models

Global spreading  
condition

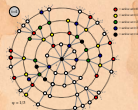
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# Time-dependent solutions

## Synchronous update

- ▶ Done: Evolution of  $\phi_t$  and  $\theta_t$  given exactly by the maps we have derived.

## Asynchronous updates

- ▶ Update nodes with probability  $\alpha$ .
- ▶ As  $\alpha \rightarrow 0$ , updates become effectively independent.
- ▶ Now can talk about  $\phi(t)$  and  $\theta(t)$ .

Contagion

Basic Contagion  
Models

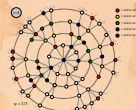
Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks

Theory

References



# References I

- [1] S. Bikhchandani, D. Hirshleifer, and I. Welch.  
A theory of fads, fashion, custom, and cultural  
change as informational cascades.  
[J. Polit. Econ.](#), 100:992–1026, 1992.
- [2] S. Bikhchandani, D. Hirshleifer, and I. Welch.  
Learning from the behavior of others: Conformity,  
fads, and informational cascades.  
[J. Econ. Perspect.](#), 12(3):151–170, 1998. [pdf](#) (田)
- [3] J. M. Carlson and J. Doyle.  
Highly optimized tolerance: A mechanism for power  
laws in design systems.  
[Phys. Rev. E](#), 60(2):1412–1427, 1999. [pdf](#) (田)

Contagion

Basic Contagion  
Models

Global spreading  
condition

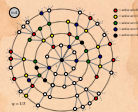
Social Contagion  
Models

Network version

All-to-all networks

Theory

References



# References II

- [4] J. M. Carlson and J. Doyle.  
Highly optimized tolerance: Robustness and design  
in complex systems.  
[Phys. Rev. Lett., 84\(11\):2529–2532, 2000.](#) pdf (田)
- [5] P. S. Dodds, K. D. Harris, and J. L. Payne.  
Physical, transparent derivation of the contagion  
condition for spreading processes on generalized  
random networks.  
<http://arxiv.org/abs/1101.5591>, 2011.
- [6] J. P. Gleeson.  
Cascades on correlated and modular random  
networks.  
[Phys. Rev. E, 77:046117, 2008.](#) pdf (田)

Contagion

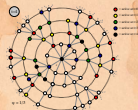
Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version  
All-to-all networks  
Theory

References





# References III

- [7] J. P. Gleeson and D. J. Cahalane.  
Seed size strongly affects cascades on random networks.  
[Phys. Rev. E, 75:056103, 2007. pdf](#) (田)
- [8] M. Granovetter.  
Threshold models of collective behavior.  
[Am. J. Sociol., 83\(6\):1420–1443, 1978. pdf](#) (田)
- [9] T. C. Schelling.  
Dynamic models of segregation.  
[J. Math. Sociol., 1:143–186, 1971.](#)
- [10] T. C. Schelling.  
Hockey helmets, concealed weapons, and daylight saving: A study of binary choices with externalities.  
[J. Conflict Resolut., 17:381–428, 1973. pdf](#) (田)

Contagion

Basic Contagion Models

Global spreading condition

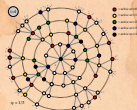
Social Contagion Models

Network version

All-to-all networks

Theory

References



# References IV

- [11] T. C. Schelling.  
Micromotives and Macrobehavior.  
Norton, New York, 1978.
- [12] D. Sornette.  
Critical Phenomena in Natural Sciences.  
Springer-Verlag, Berlin, 2nd edition, 2003.
- [13] D. J. Watts.  
A simple model of global cascades on random  
networks.  
Proc. Natl. Acad. Sci., 99(9):5766–5771, 2002.  
[pdf](#) (田)

Contagion

Basic Contagion  
Models

Global spreading  
condition

Social Contagion  
Models

Network version

All-to-all networks

Theory

References

