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# Complex Contagions on Configuration Model Graphs with a Power-Law Degree Distribution

Grant Schoenebeck, **Fang-Yi Yu**

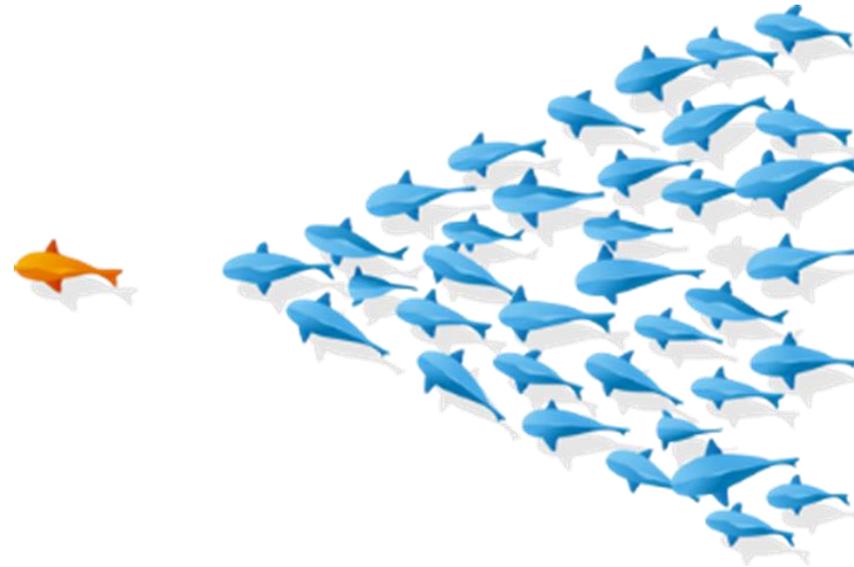


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# Contagions, diffusion, cascade...

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- Ideas, beliefs, behaviors, and technology adoption spread through network
- Why do we need to study this phenomena?
  - Better Understanding
  - Promoting good behaviors/beliefs
  - Stopping bad behavior



# Outline

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- Models
    - Complex contagions model
    - Power-Law and configuration model graph
  - Main result
  - Related work
  - A happy proof sketch
-

# Outline

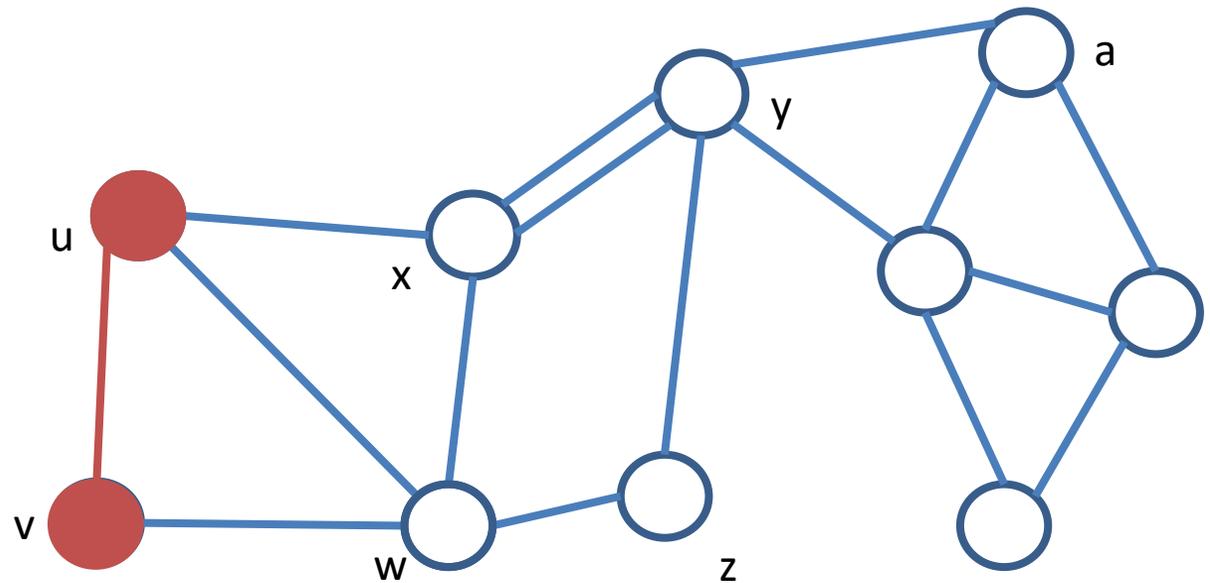
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# Model of Contagions

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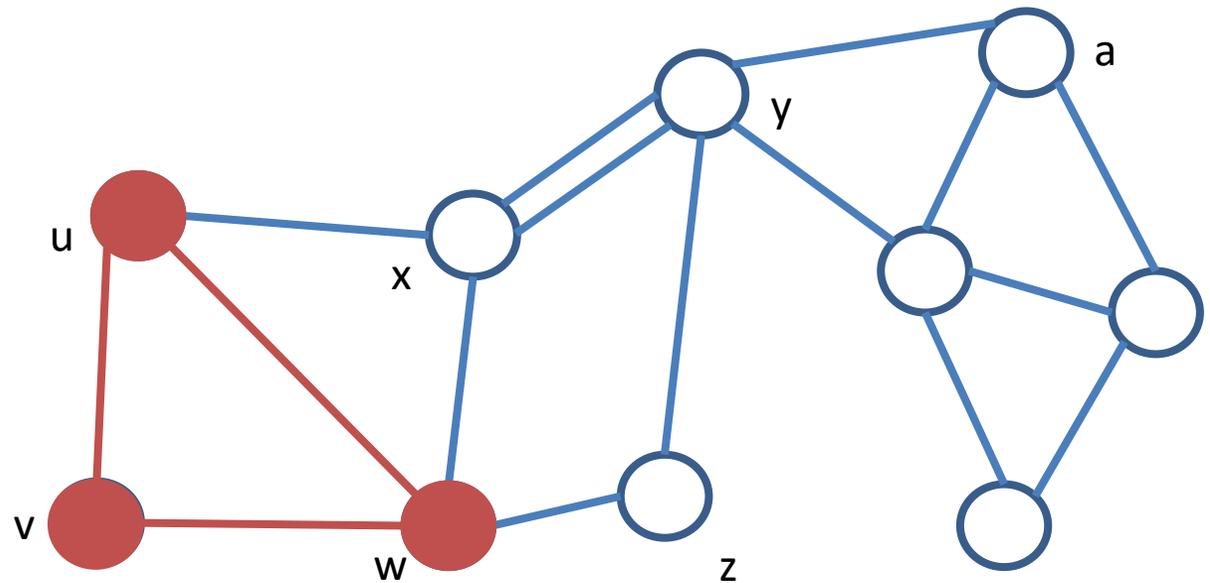
- K-Complex Contagions [GEG<sub>13</sub>; CLR 79]
  - Given an **initial infected set**  $I = \{u, v\}$ .



# Model of Contagions

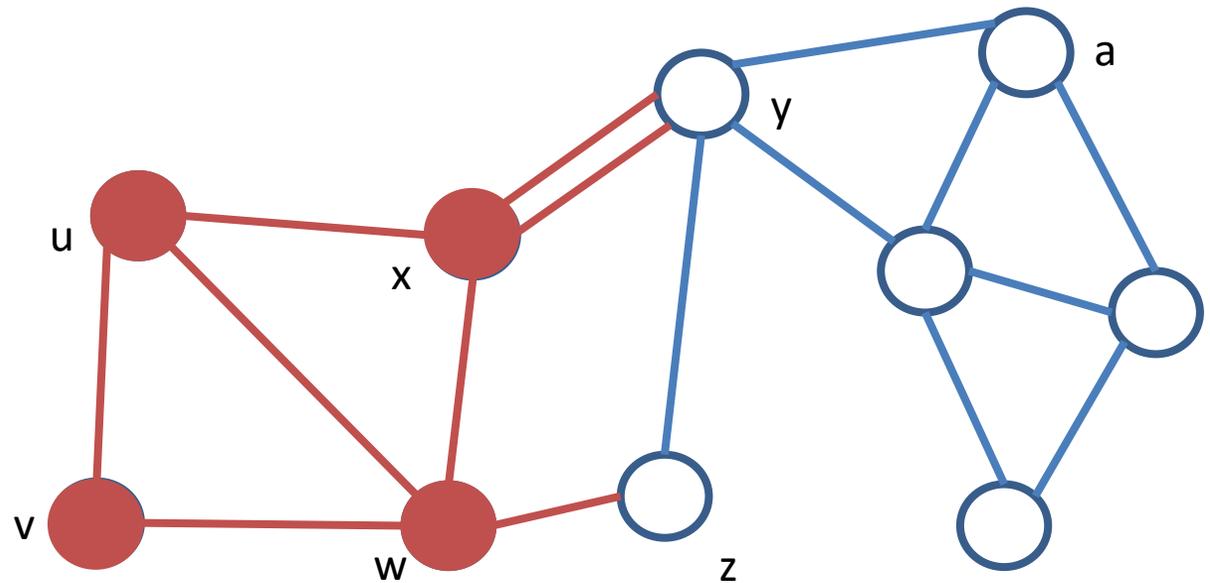
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- K-Complex Contagions [GEG<sub>13</sub>; CLR 79]
  - Given an **initial infected set**  $I = \{u, v\}$ .
  - Node becomes infected if it has at least **k** infected neighbor



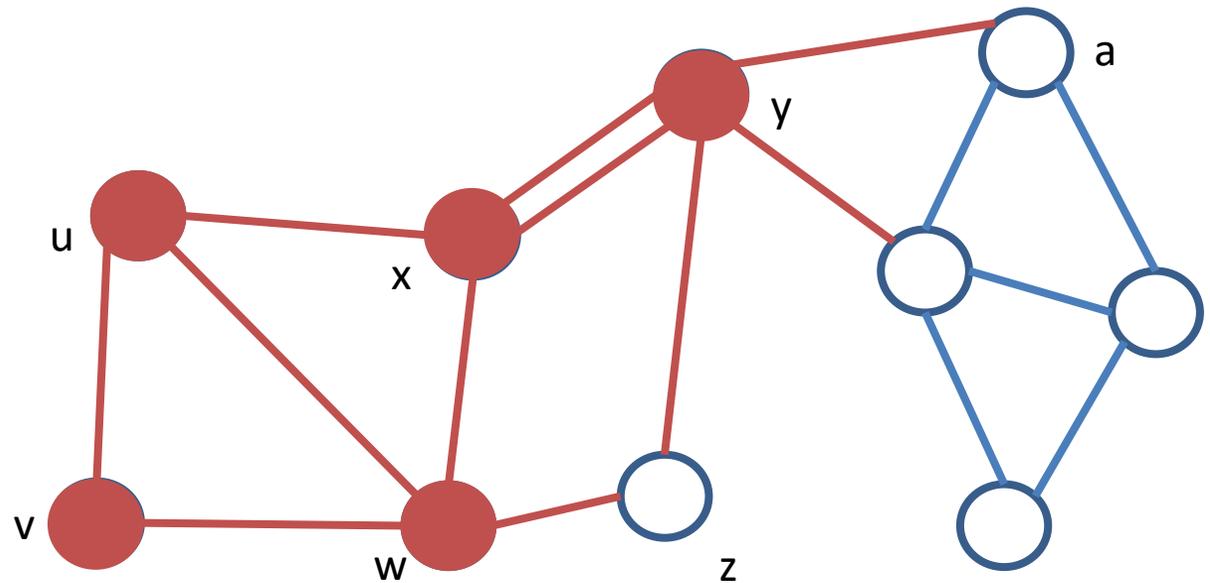
# Model of Contagions

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# Model of Contagions

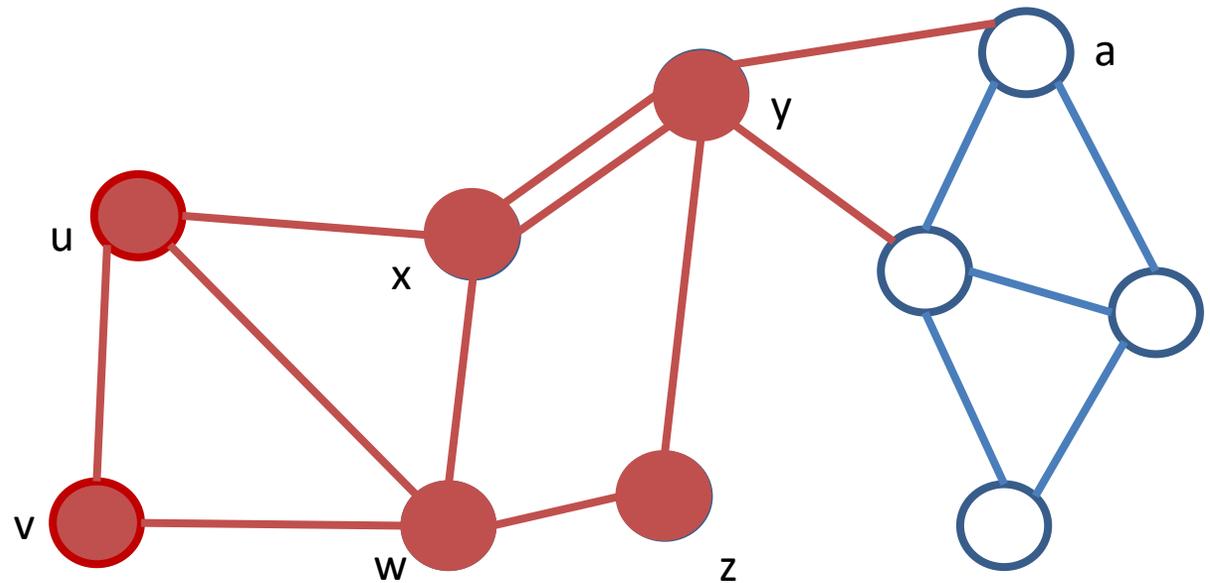
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# Model of Contagions

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  - Given an **initial infected set**  $I = \{u, v\}$ .
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# Why $k$ complex contagions?

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- One of most classical and simple contagions model
    - Threshold model [Gra 78]
    - Bootstrap percolation [CLR 79]
  - Non-submodular
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# Motivating Question

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- Do  $k$  complex contagions spread on social networks?
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# Question

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- Do  $k$  complex contagions spread on Erdos-Renyi model  $G_{n,p}$  where  $p = O\left(\frac{1}{n}\right)$ ?
    - $n$  vertices
    - Each edge  $(u, v)$  occurs with probability  $p$
  - Need  $\Omega(n)$  (random) seeds to infect constant fraction of the graph[JLTV89]?
  - Can we categorize all networks which spread slowly/quickly?
-

# What is a social network?

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- Qualitatively: special structure
    - Power law degree distribution
    - low-diameter/small-world...
  - Quantitatively: generative model?
    - Configuration model graphs
    - Preferential attachment model
    - Kleinberg's small world model
-

# Motivating Question

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- Do  $k$  complex contagions spread on social networks?
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# Motivating Question

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- Do  $k$  complex contagions spread on social networks?
    - What properties are shared by social networks?
    - Do these properties alone permit complex contagion spreads?
-

# Outline

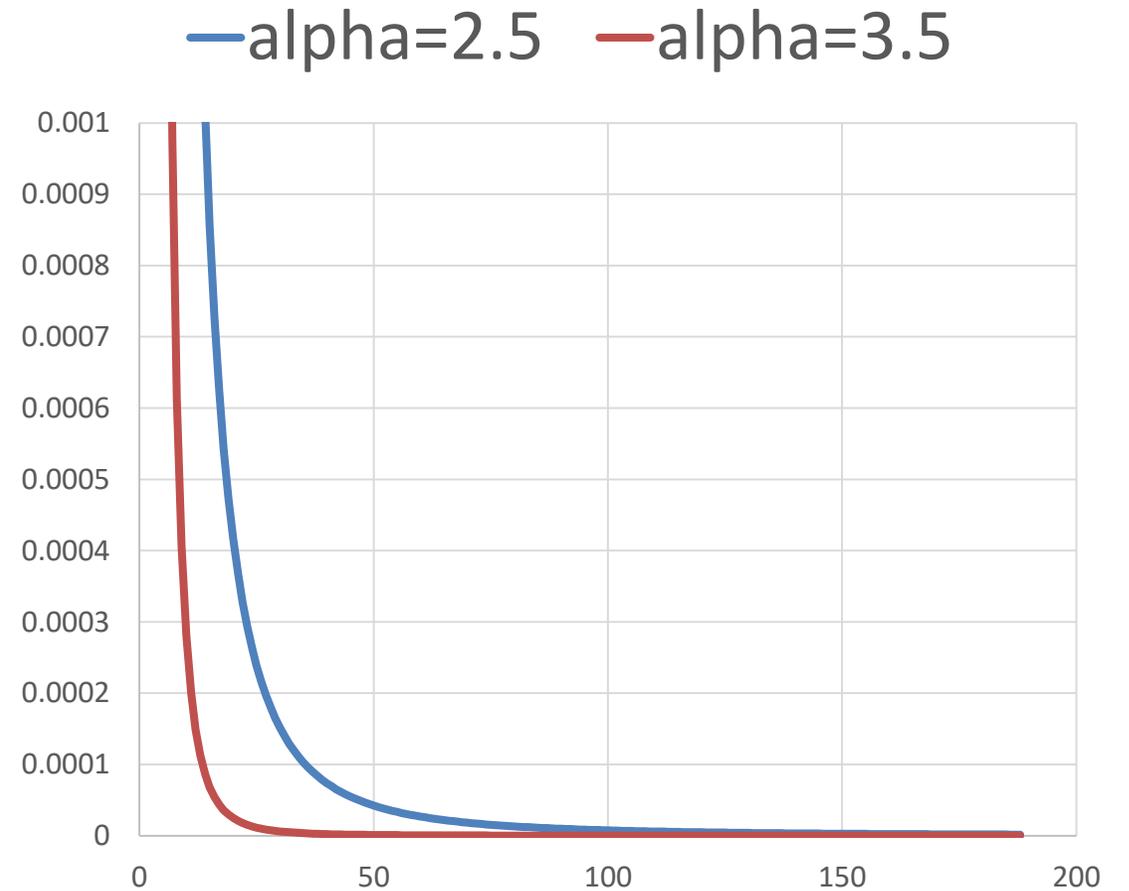
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-

# Power-law distribution

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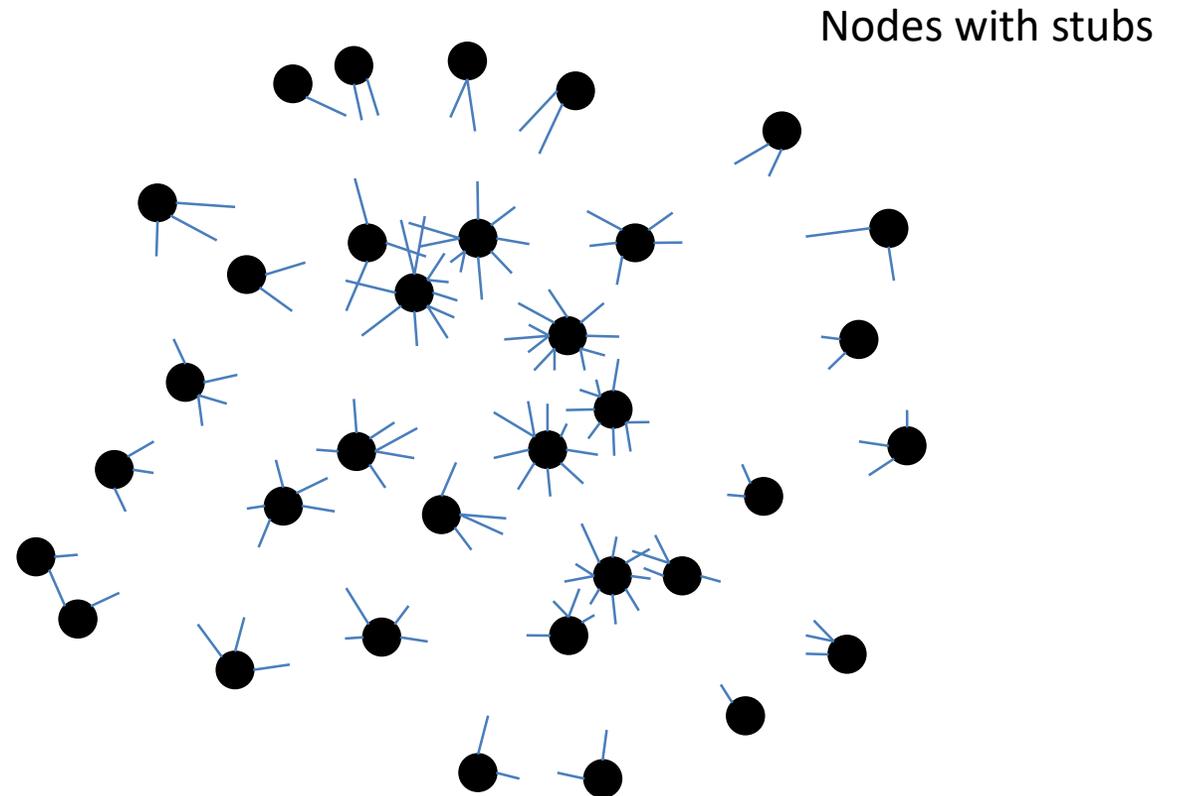
- A power-law distribution with  $\alpha$  if the  $\Pr[X = x] \sim x^{-\alpha}$



# Configuration Model

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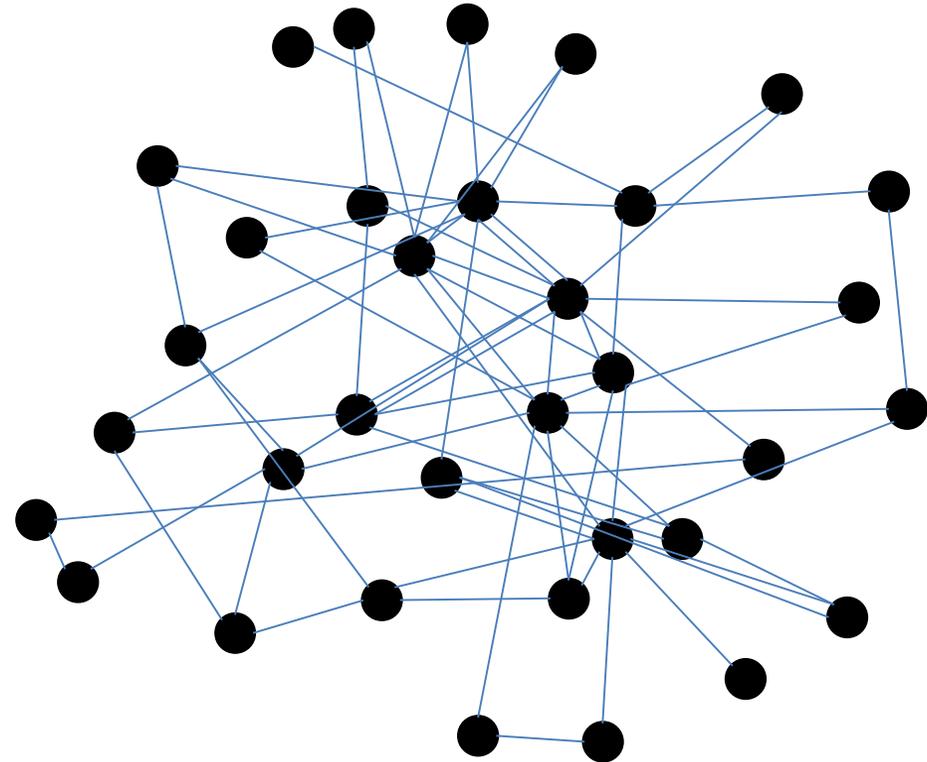
- Given a degree sequence  $\deg(v_1), \deg(v_2), \dots, \deg(v_n)$
- The node  $v_i$  has  $\deg(v_i)$  stubs



# Configuration Model

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- Given a **degree sequence**  $\deg(v_1), \deg(v_2), \dots, \deg(v_n)$
- The node  $v_i$  has  $\deg(v_i)$  **stubs**
- Choose a uniformly random matching on the **stubs**



# Outline

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- Models
    - Complex contagions model
    - Power-Law and configuration model graph
  - **Main result**
  - Related work
  - A happy proof sketch
-

# Theorems

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## Main result

- Configuration Model
  - power-law degree distribution  $2 < \alpha < 3$
- Initial infected node
  - the highest degree node
- $k$ -complex contagions spreads to  $\Omega(1)$  fraction of nodes with high probability.

## Corollary from [Amini 10]

- Configuration Model
    - power-law degree distribution  $3 < \alpha$
  - Initial infected nodes:
    - $o(1)$  fraction of highest degree node
  - $k$ -complex contagions spreads to  $o(1)$  fraction of nodes with high probability.
-

# The Bottom Line

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## Main result

- Configuration Model
  - power-law degree distribution  $2 < \alpha < 3$
- $k$ -complex contagions
  - Initial infected node: the highest degree node
- Contagions spreads to  $\Omega(1)$  fraction of nodes with high probability.

## Corollary from [Amini 10]

- Configuration Model
  - power-law degree distribution  $3 < \alpha$
- $k$ -complex contagions
  - Initial infected node:  $o(1)$  fraction of highest degree node
- Contagions spreads to  $o(1)$  fraction of nodes with high probability.

Complex contagions spread on most of graphs with power-law degree distribution  $2 < \alpha < 3$

# Outline

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- Models
    - Complex contagions model
    - Power-Law and configuration model graph
  - Main result
  - **Related work**
  - A happy proof sketch
-

# What has been done?

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|             |                          |  |                                     |
|-------------|--------------------------|--|-------------------------------------|
| Lattice[98] | Random<br>regular[07]    | Configuration<br>model $\alpha > 3$ [10] | Watts-Strogatz[13]<br>Kleinberg[14] |
| Tree[79,06] | Erdos-Renyi<br>model[12] | Chung-Lu<br>model[12]                    | Preferential<br>Attachment[14]      |

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# Start with $G_{n,p}$

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|             |                       |  |                                     |
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# k-complex contagions don't spread

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## Physics

Lattice[98]

Random  
regular[07]

Configuration  
model  $\alpha > 3$  [10]  
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model  $2 < \alpha < 3$  [16]

Watts-Strogatz[13]  
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# k-complex contagions spread

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## Network Science

|             |                       |  |                                     |
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# Things get complicated

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Physics

Network Science

Lattice[98]

Tree[79,06]

Random  
regular[07]

Erdos-Renyi  
model[12]

Configuration  
model  $\alpha > 3$  [10]  
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# What have we learned?

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Physics

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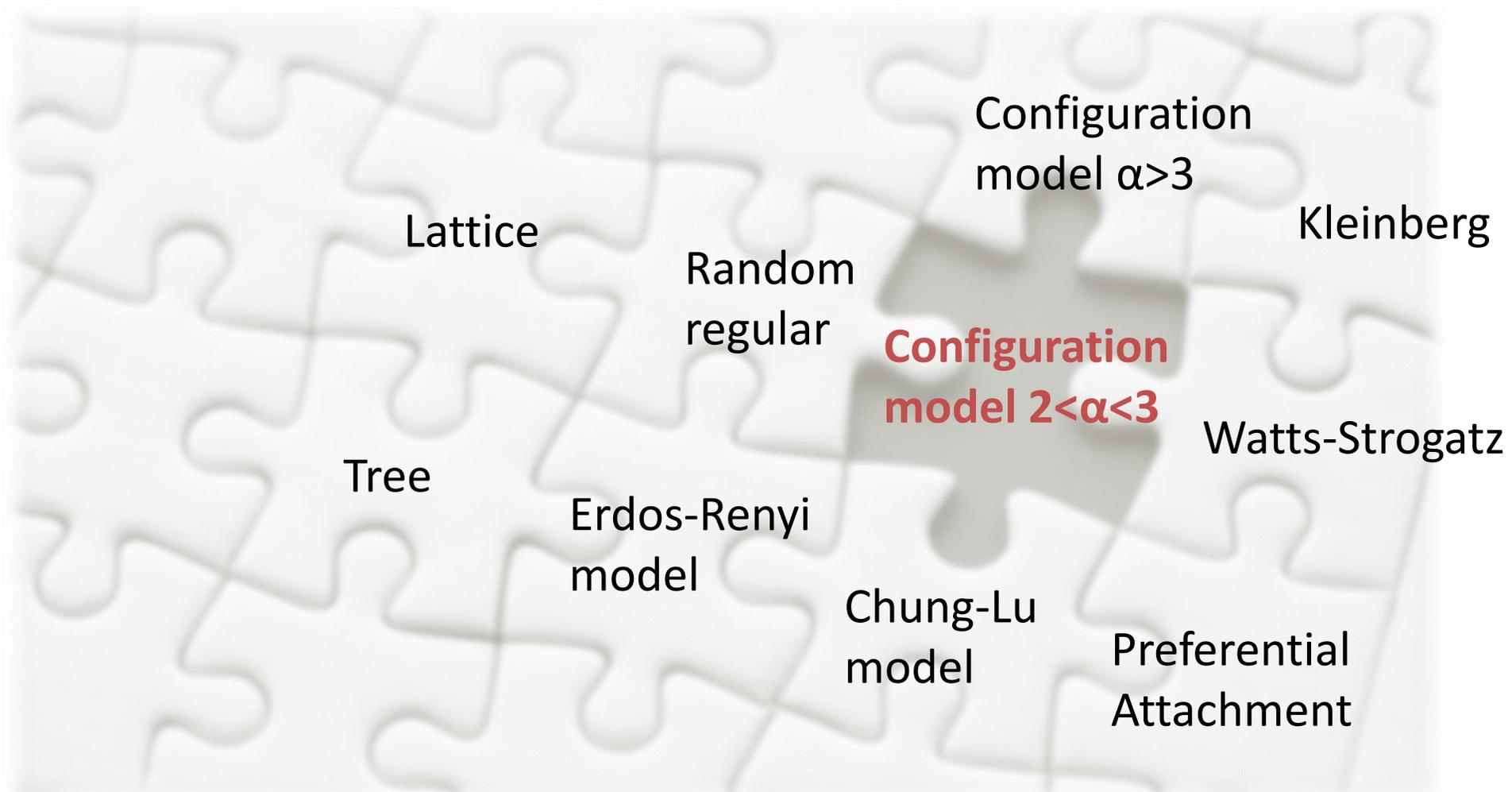
Chung-Lu  
model[12]

Preferential  
Attachment[14]

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# Why do we want to solve it?

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# Outline

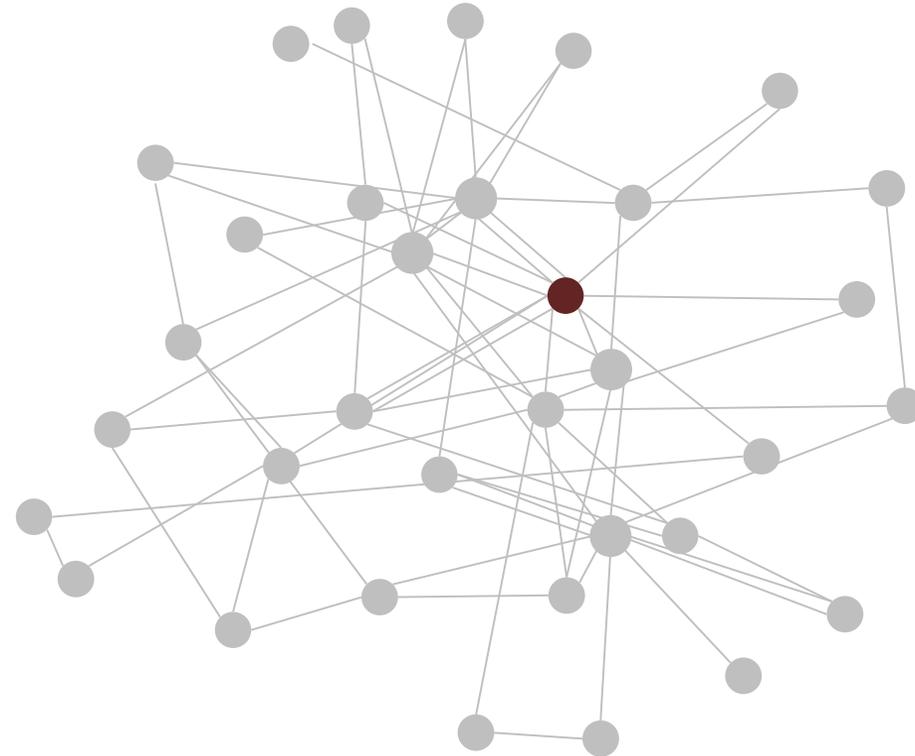
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- Models
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# Idea

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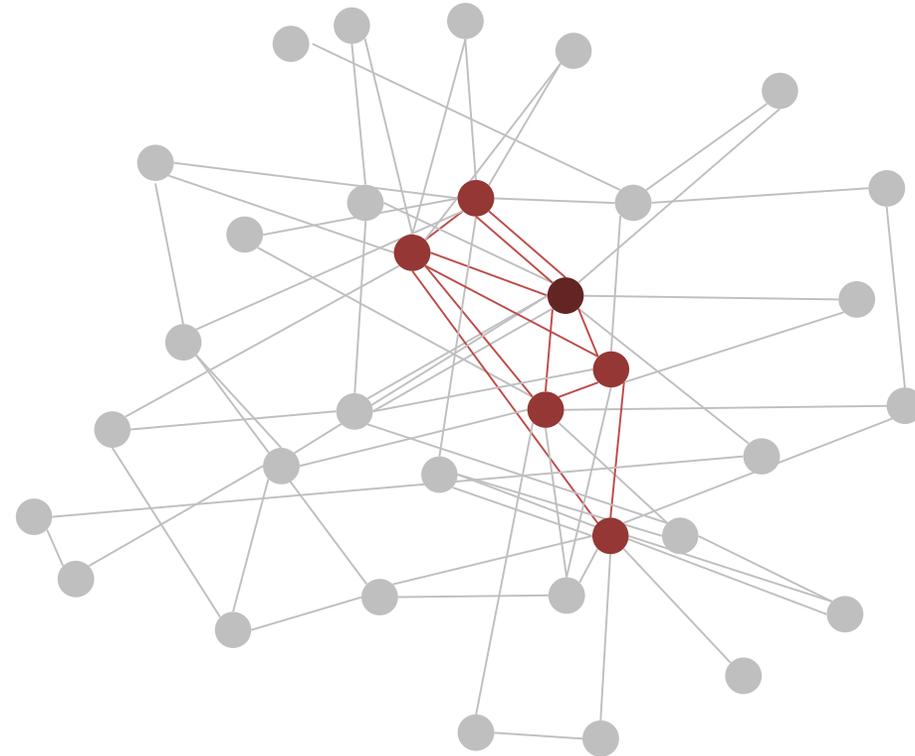
- Restrict contagions from high degree node to low degree nodes
- Reveal the edges when needed



# Observations

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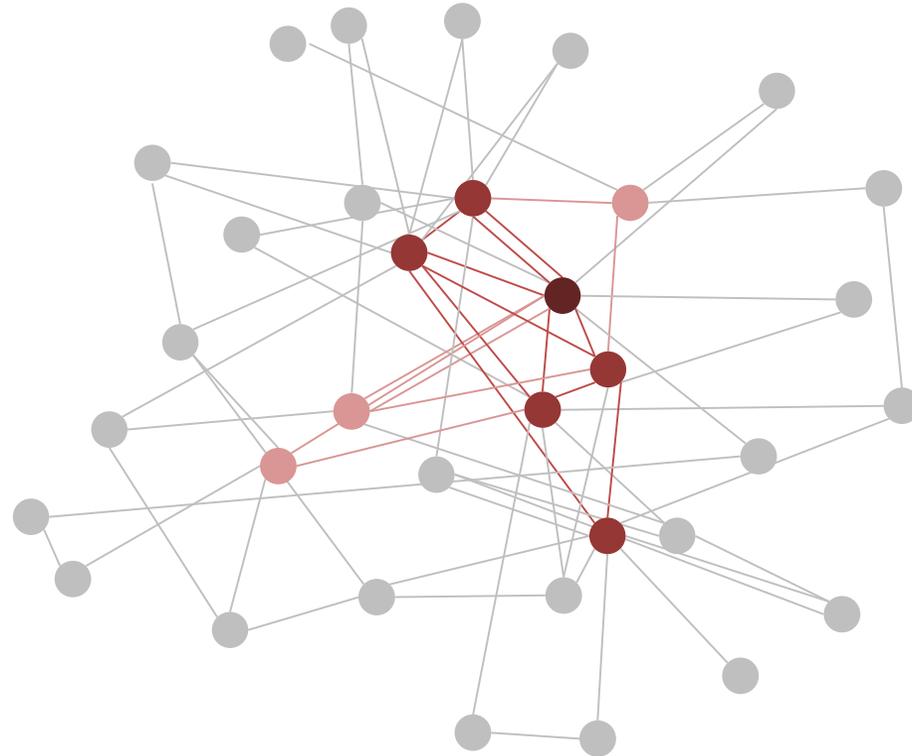
- The highest degree nodes forms clique



# Observations

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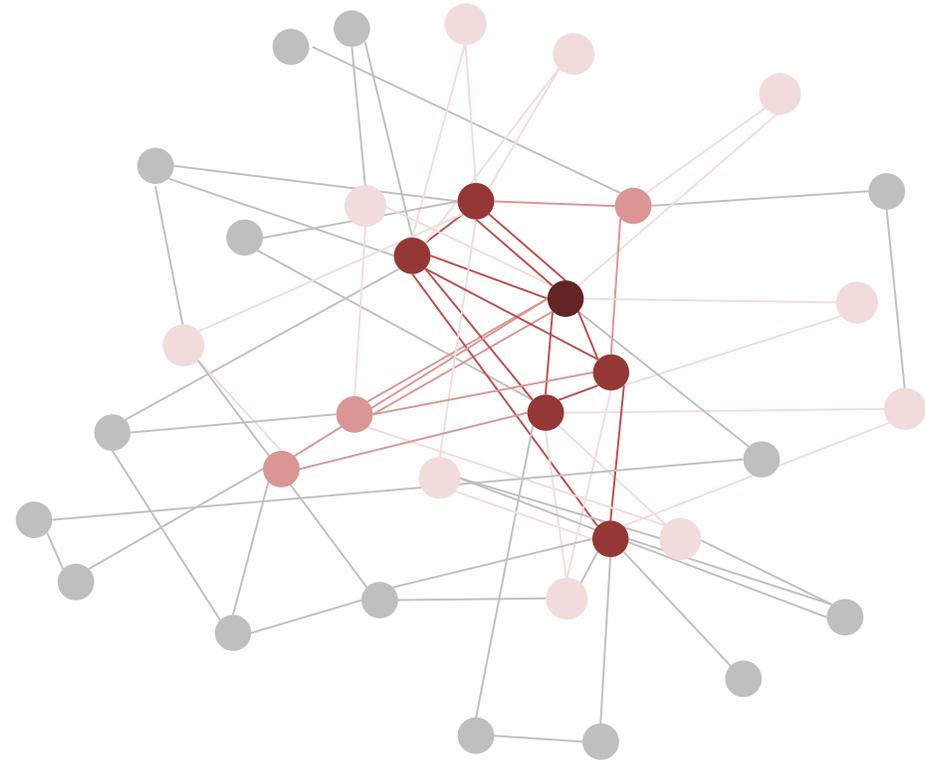
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- $k$  degree node has many edges to  $l$  degree nodes where  $l > k$



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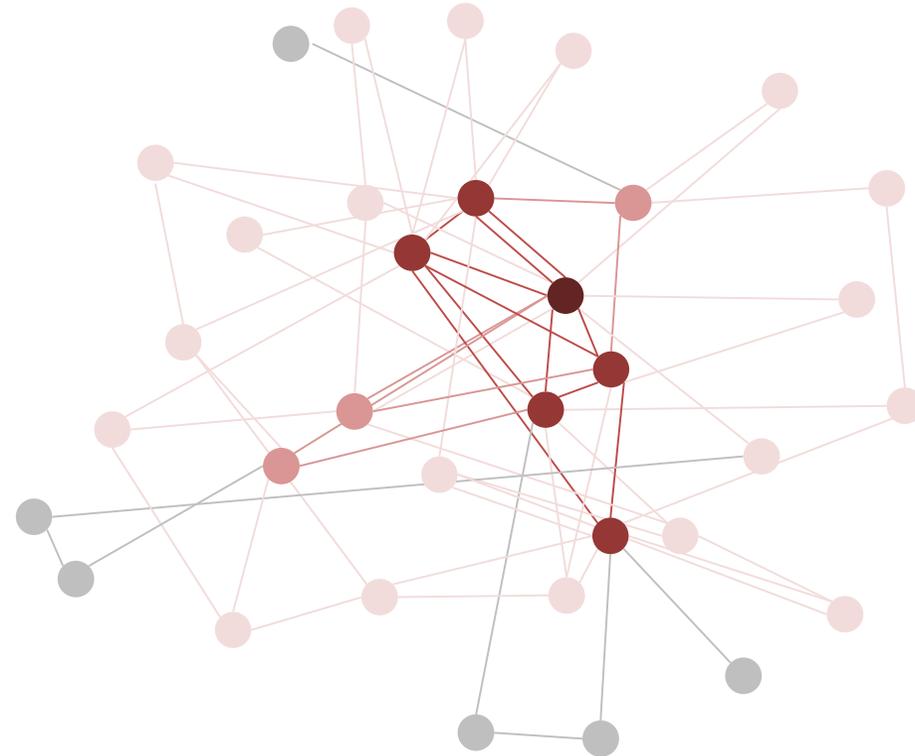
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- Inductive structure



# Observations

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- The highest degree nodes forms clique
- $k$  degree node has many edges to  $l$  degree nodes where  $l > k$
- Inductive structure



# Previous tool and challenges

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Physics

Network Science

Lattice[98]

Random  
regular[07]

Configuration  
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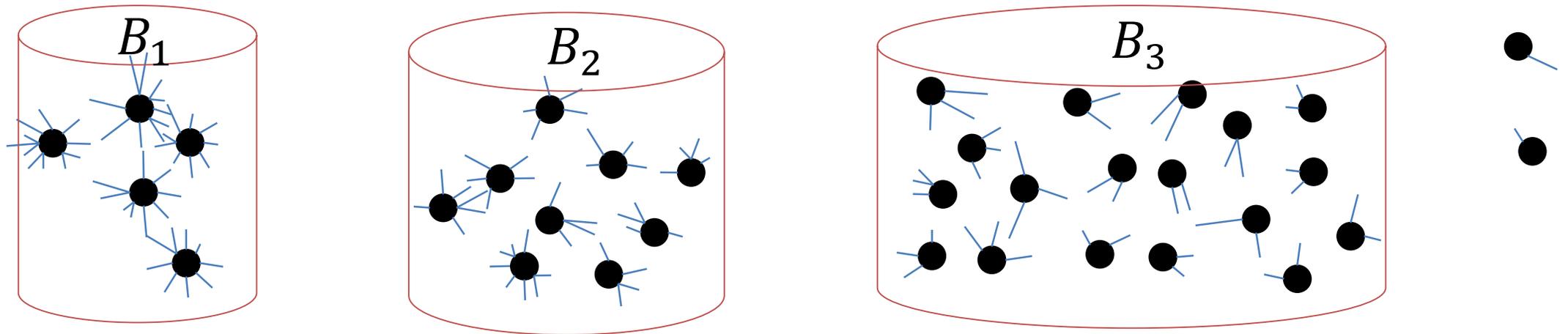
# Inductive Structure

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- Partition nodes into buckets ordered by degree of nodes

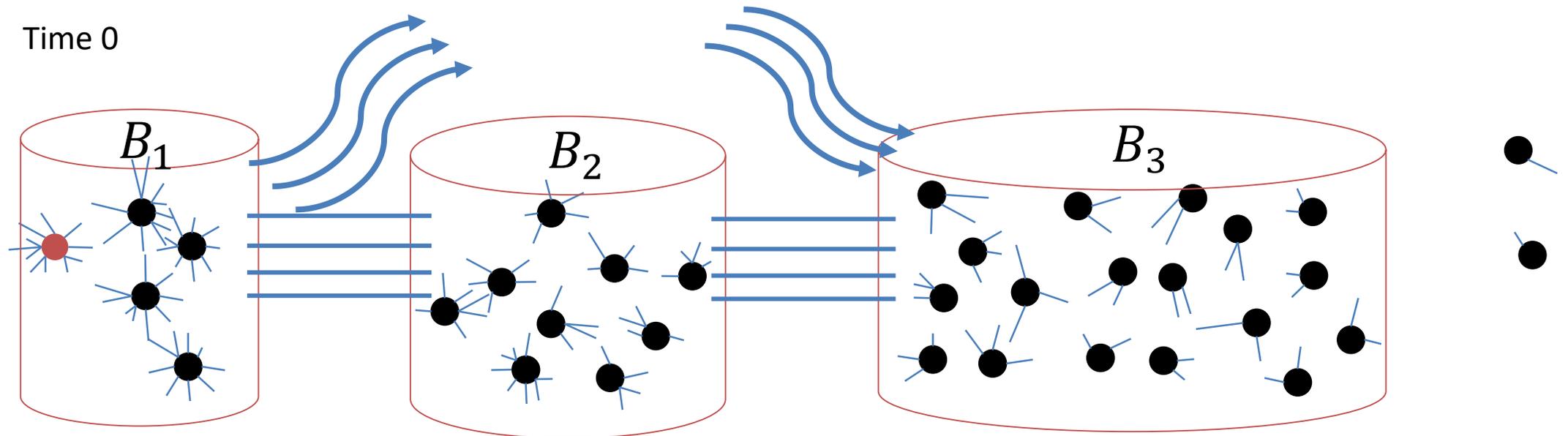
$$B_1, B_2, \dots, B_l$$

- Induction: if infection spreads on previous buckets  $B_i$  where  $i < k$ , the infection also spread on bucket  $B_k$ .



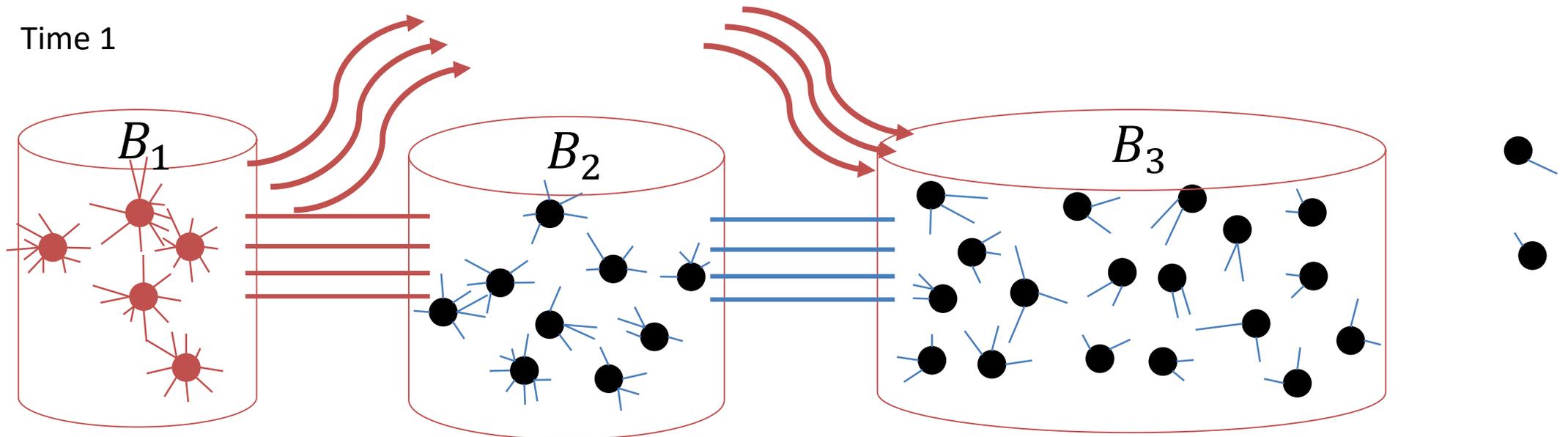
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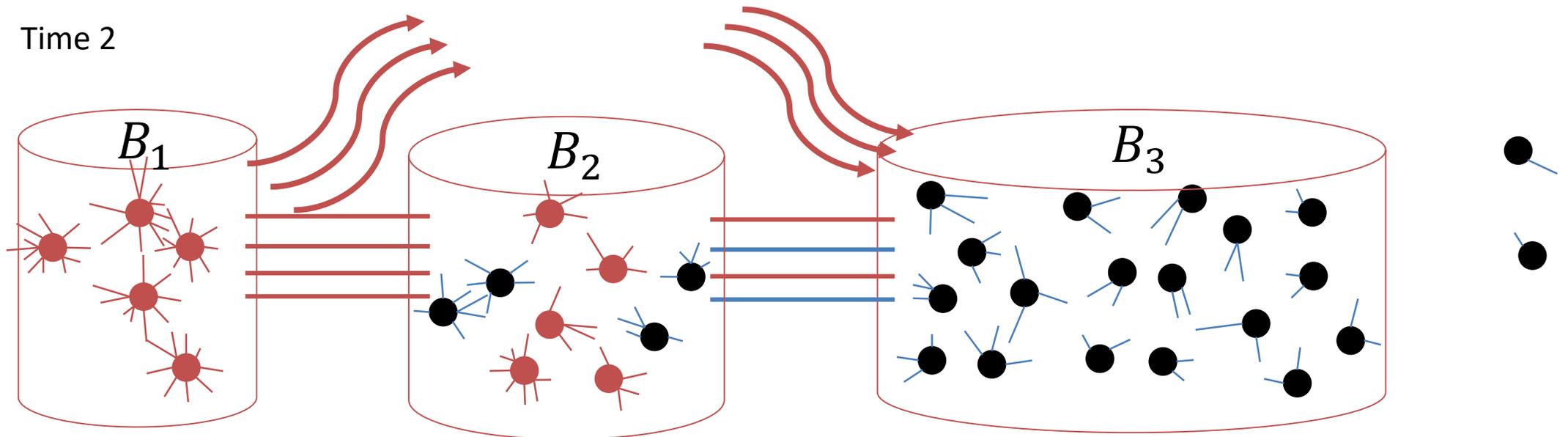
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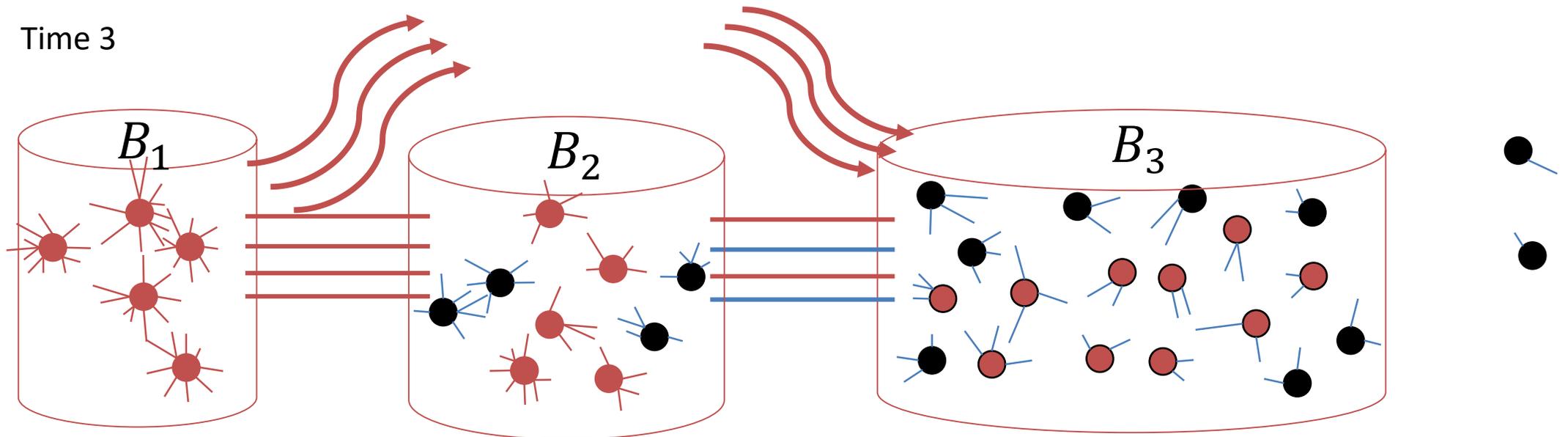
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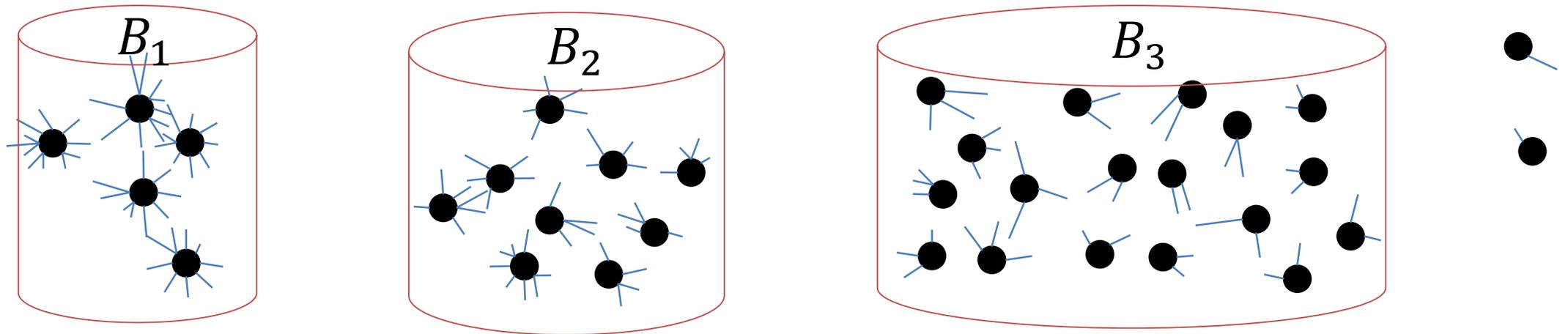
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# Inductive Structure

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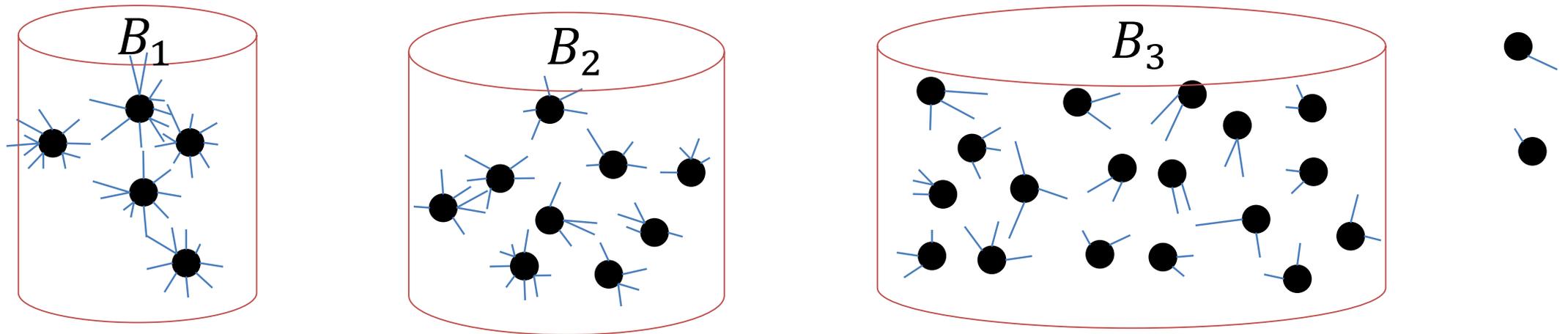
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  - Well connection between buckets
  - Infection spread in buckets



# Inductive Structure

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- Induction: if infection spreads on previous buckets  $B_i$  where  $i < k$ , the infection also spread on bucket  $B_k$ .
  - Well connection between buckets: Chernoff bound
  - Infection spread in buckets: Chebeshev's inequality



# Thanks for your listening

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# How do we solve it?

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- Chebyshev's inequality

$$\Pr[Z > E[Z] + t] \leq \frac{\text{Var}[Z]}{t^2}$$

- Chernoff-type bound

$$\Pr[Z_n > Z_0 + t] \leq \exp\left(\frac{-t^2}{c}\right)$$

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