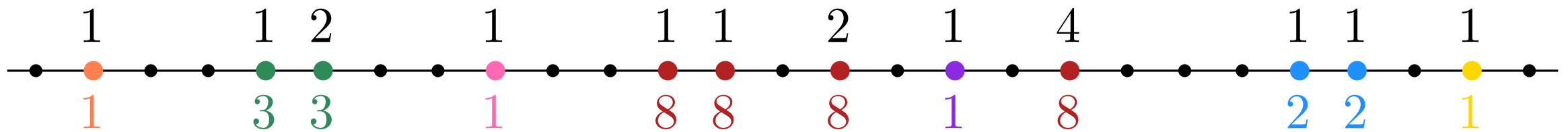
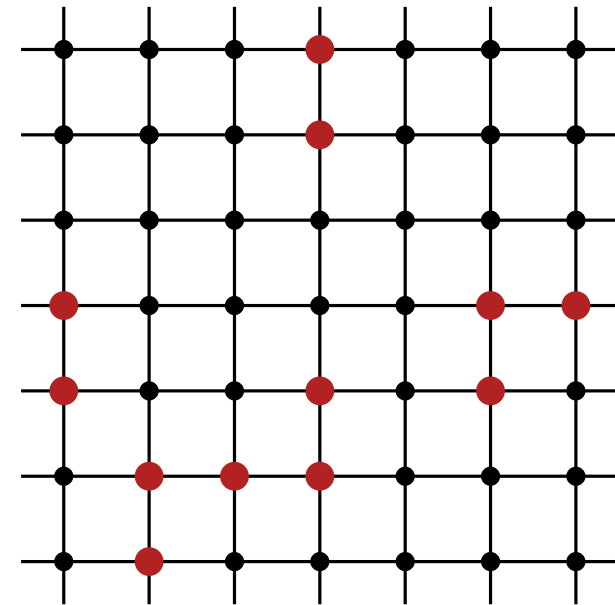
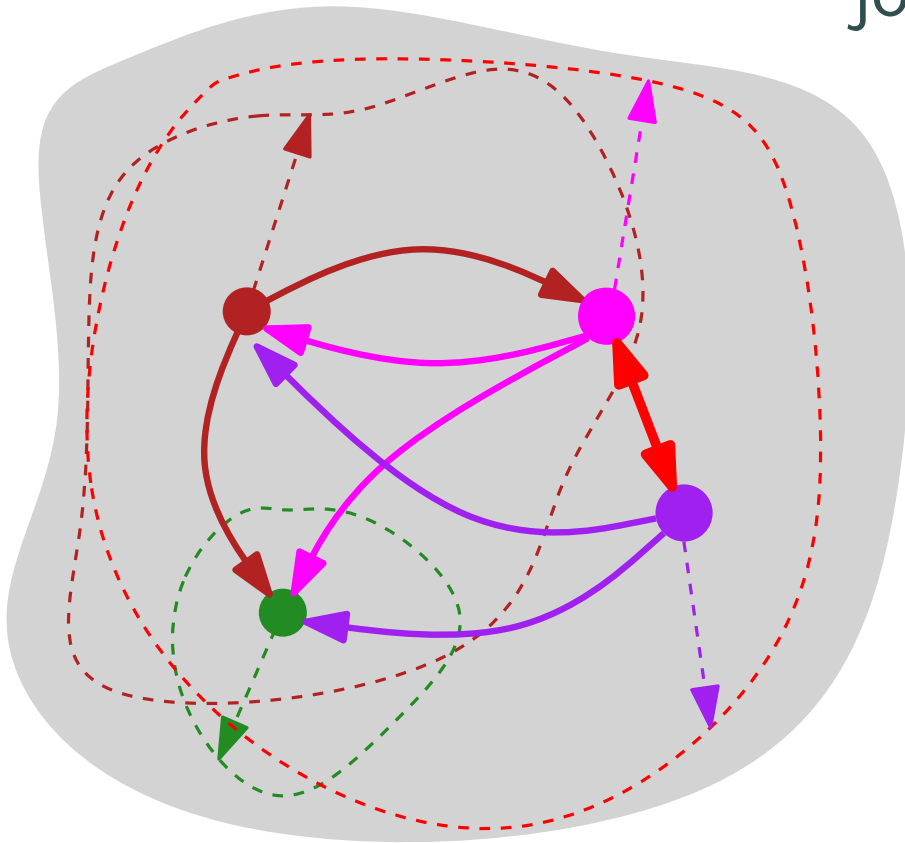


# Percolation by cumulative merging and phase transition of the contact process

Laurent Ménard (Paris Nanterre)

joint work with **Arvind Singh** (Paris Orsay)

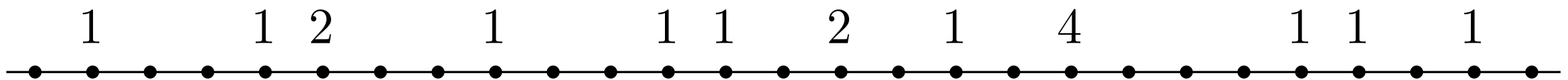


# Outline

1. Cumulative merging
2. Phase transition for cumulative merging percolation
3. The contact process
4. Heuristics for the contact process
5. Link with cumulative merging

# Cumulative Merging?

Take  $G = (V, E, r)$  any locally finite connected weighted graph with  $r : V \rightarrow [0, +\infty]$

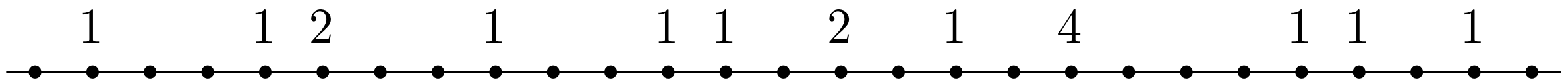


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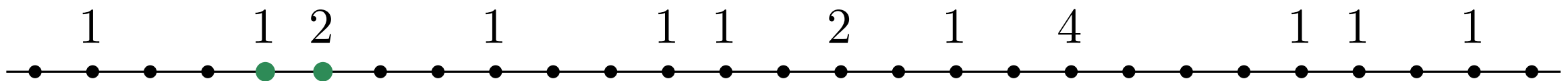


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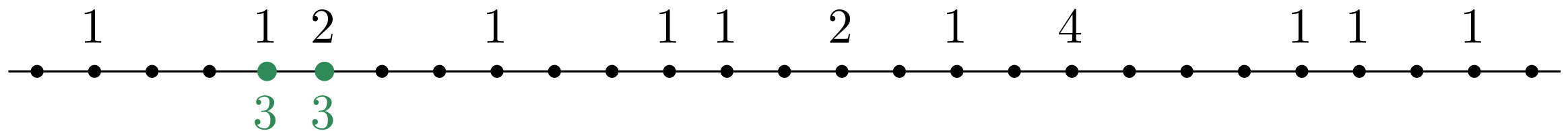


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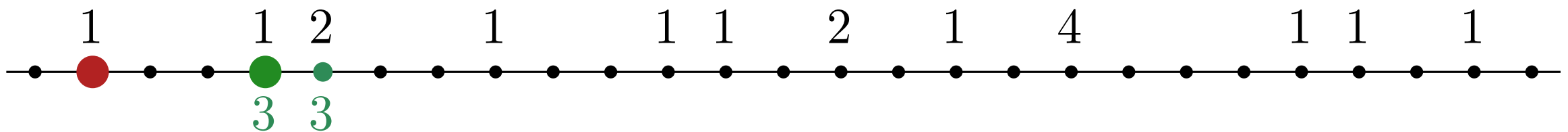


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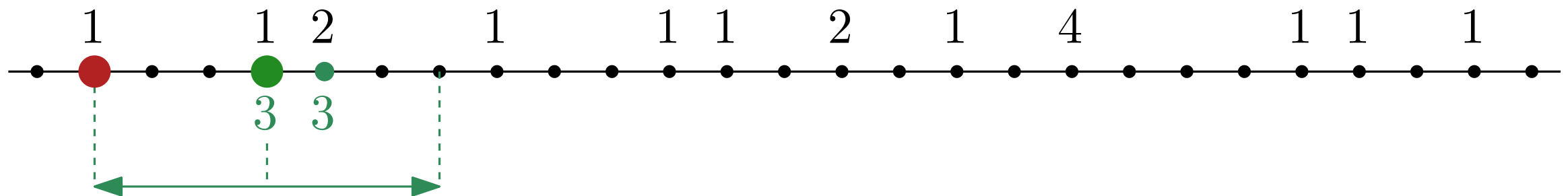


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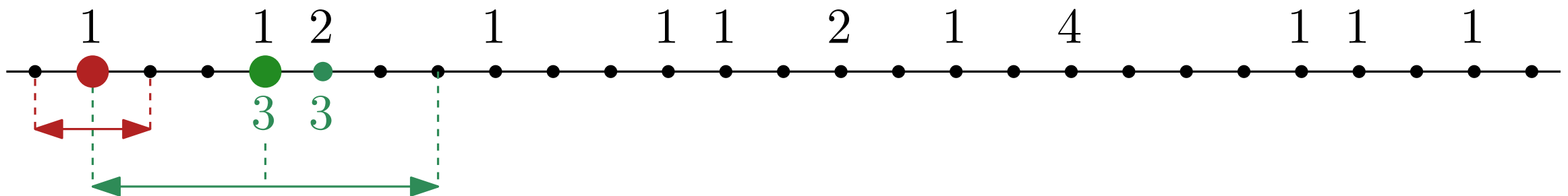


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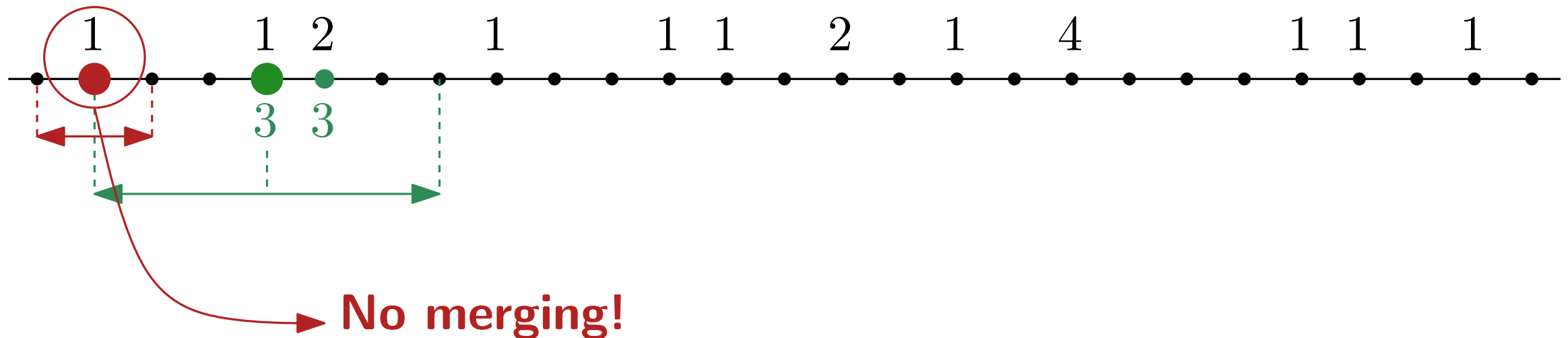


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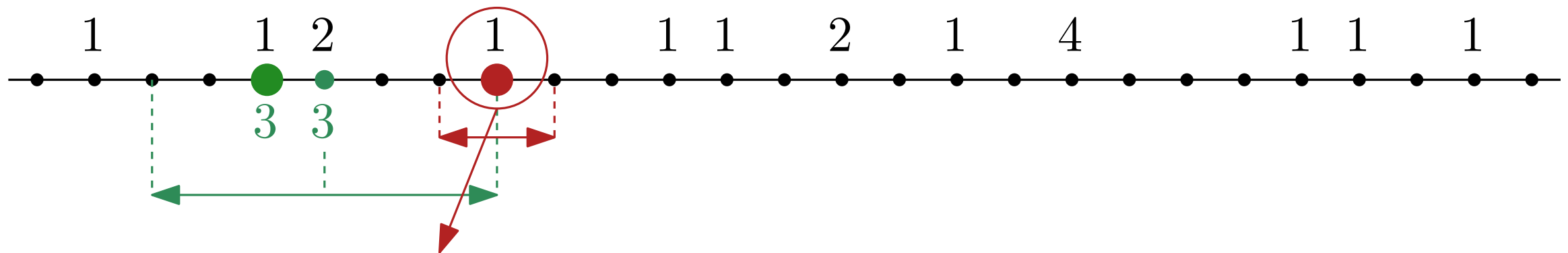


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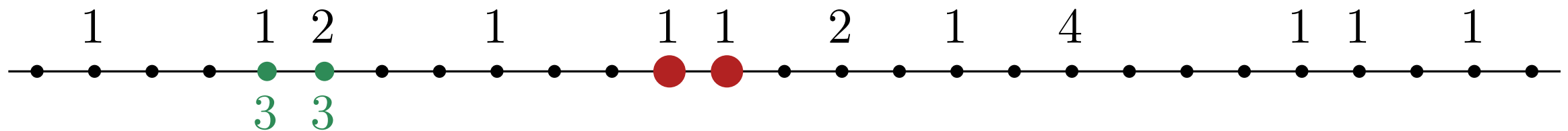
**No merging!**

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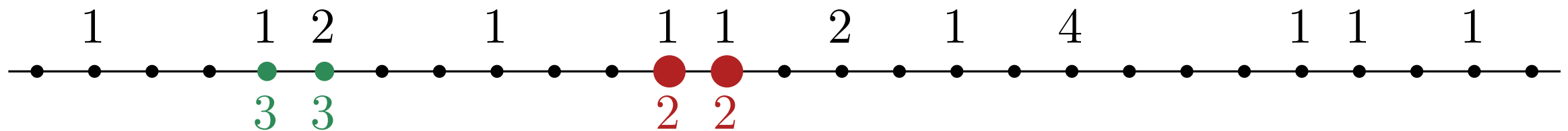


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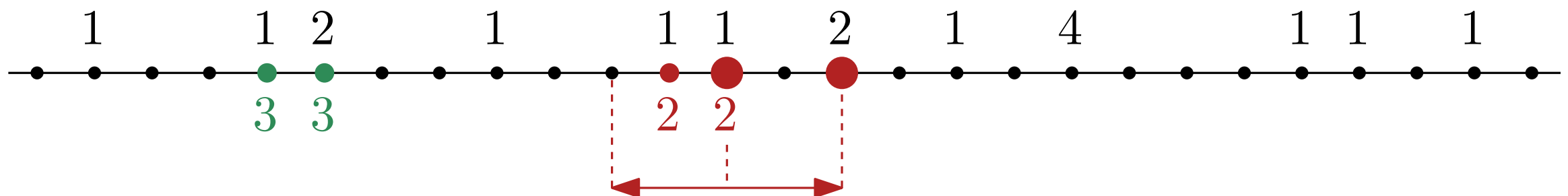


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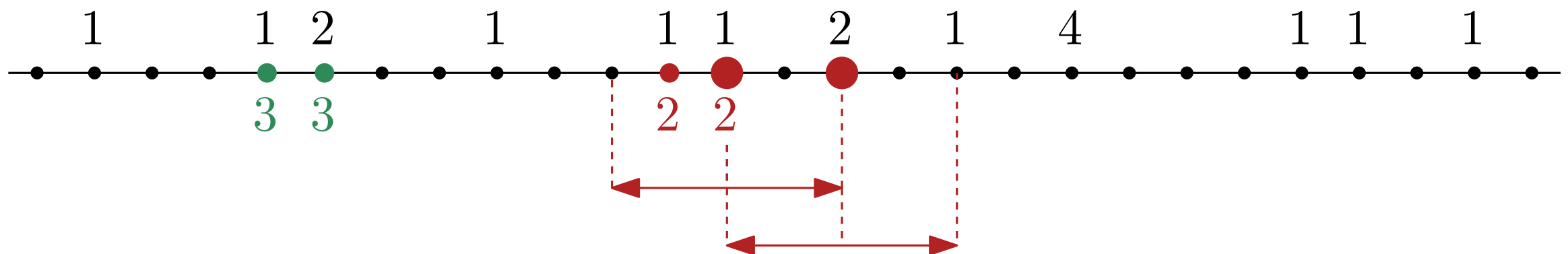


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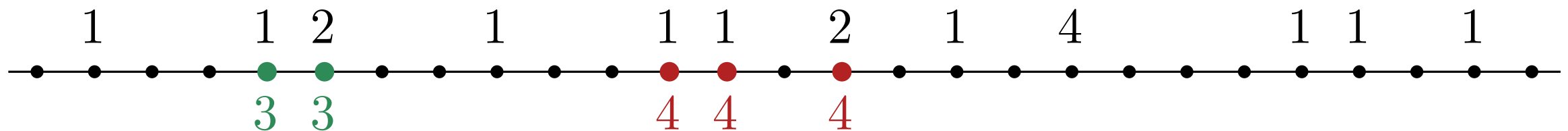


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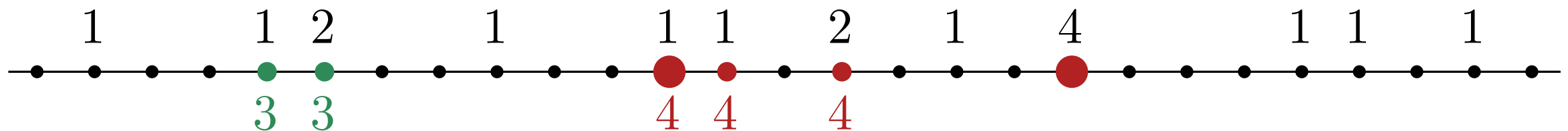


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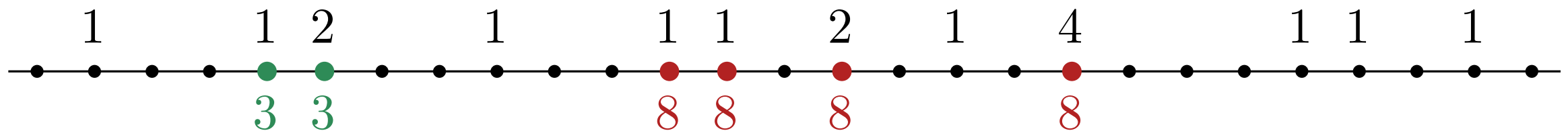


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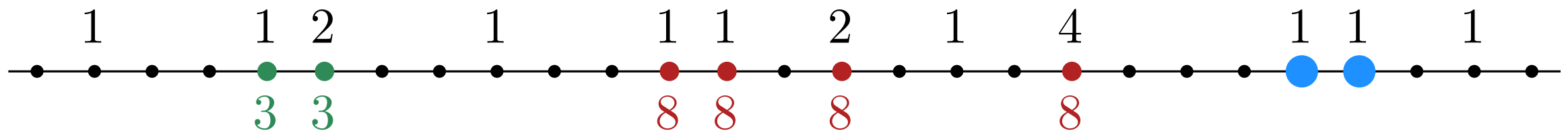


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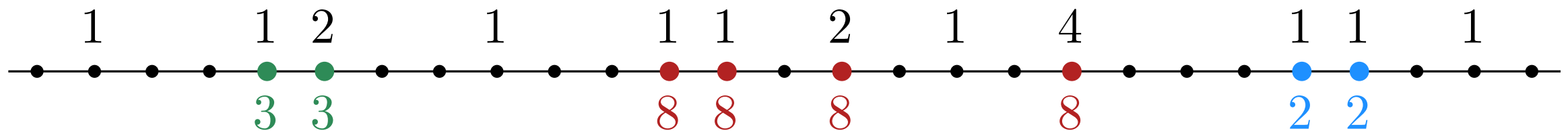


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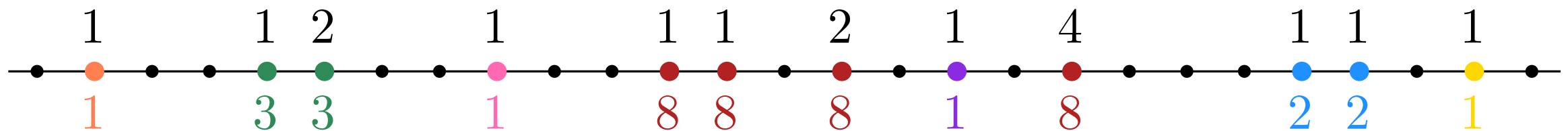


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## Definition

a partition  $\mathcal{P}$  of  $V$  is **admissible** iff  $\forall A \neq B \in \mathcal{P}$ :

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## Definition

$$\mathcal{C}(G, r) := \bigwedge_{\text{admissible } \mathcal{P}} \mathcal{P} \quad (\text{finest admissible partition})$$



# Merging operators

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For  $x \neq y \in V$ ,  $M_{x,y} : \{\text{partitions of } V\} \rightarrow \{\text{partitions of } V\}$   
defined by

$$M_{x,y}(\mathcal{P}) := \begin{cases} (\mathcal{P} \setminus \{\mathcal{P}_x, \mathcal{P}_y\}) \cup \{\mathcal{P}_x \cup \mathcal{P}_y\} & \text{if } \mathcal{P}_x \neq \mathcal{P}_y \text{ and} \\ & d(x, y) \leq r(\mathcal{P}_x) \wedge r(\mathcal{P}_y), \\ \mathcal{P} & \text{otherwise.} \end{cases}$$

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## Proposition:

The merging operators are monotone: for every  $x \neq y \in V$  and every partitions  $\mathcal{P}$  and  $\mathcal{P}'$

- $\mathcal{P}$  is finer than  $M_{x,y}(\mathcal{P})$ ;
- If  $\mathcal{P}$  is finer than  $\mathcal{P}'$ , then  $M_{x,y}(\mathcal{P})$  is finer than  $M_{x,y}(\mathcal{P}')$ .

# Cumulative Merging

## Proposition:

Take  $(x_n, y_n) \in V^{\mathbb{N}} \times V^{\mathbb{N}}$  such that for every  $x \neq y \in V$ :

$$\{x_n, y_n\} = \{x, y\} \text{ for infinitely many } n.$$

Then

$$\mathcal{C}(V, E, r) = \lim_{n \rightarrow \infty} \uparrow M_{x_n, y_n} \circ \cdots \circ M_{x_1, y_1}(\bar{V})$$

where  $\bar{V} = \{\{x\}\}_{x \in V}$  is the finest partition of  $V$ .

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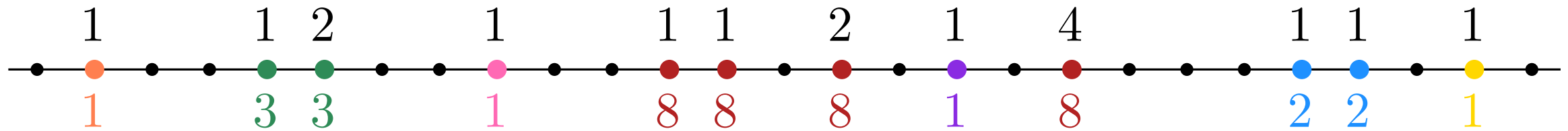
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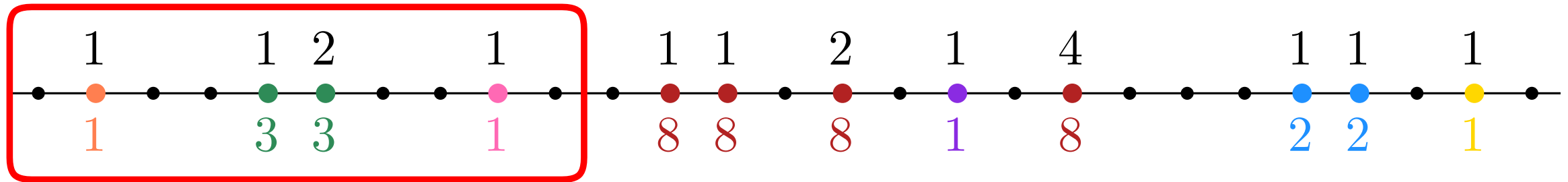
**Proof:** Monotonicity of the merging operators.

# Cumulative Merging: Basic observations



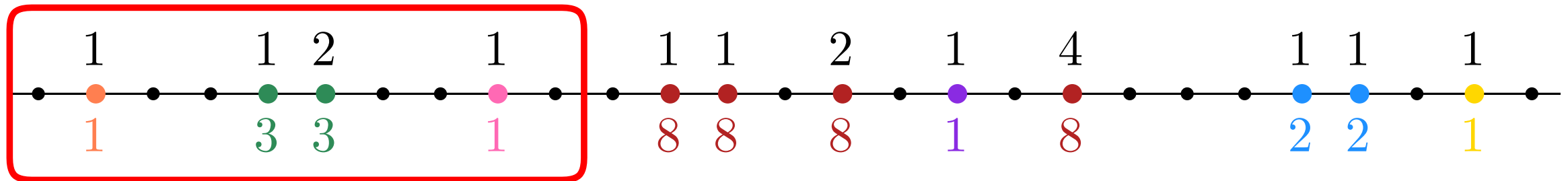
- Clusters in  $\mathcal{C}$  are not necessarily connected sets!
- If  $r(x) < 1$ , then  $\{x\} \in \mathcal{C}$ .
- If  $\mathcal{C}$  has an infinite cluster, it has infinite weight and is unique.
- For any  $C \in \mathcal{C}$ , one has  $|C| \leq \max\{1, r(C)\}$ .

# CMP: Stable sets



Vertices inside that box will never merge again.

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Vertices inside that box will never merge again.

## Definition:

Fix  $H \subset V$ . We say that  $H$  is a **stable set** iff:

$$\forall C \in \mathcal{C}(H, E_H, r) \text{ one has } B(C, r(C)) \subset H.$$

## Remark:

- Unions and intersections of stable sets are stable.
- Being stable is a **local** property.
- If  $H$  is stable, then

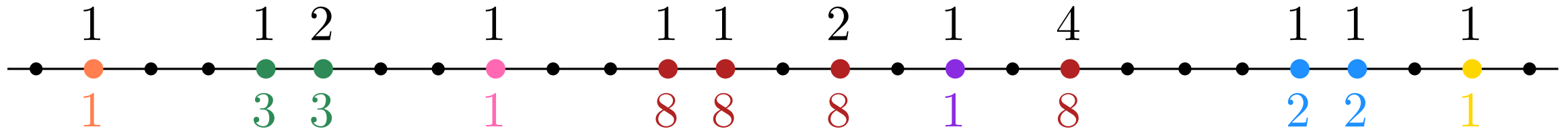
$$\mathcal{C}(G) = \mathcal{C}(H) \sqcup \mathcal{C}(G \setminus H).$$



# CMP: Stabilisers

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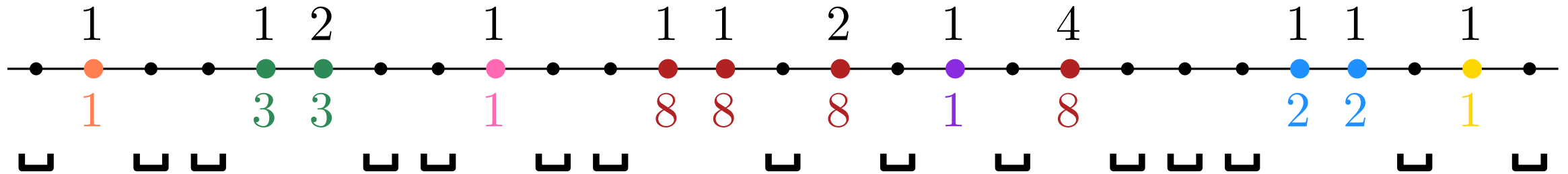
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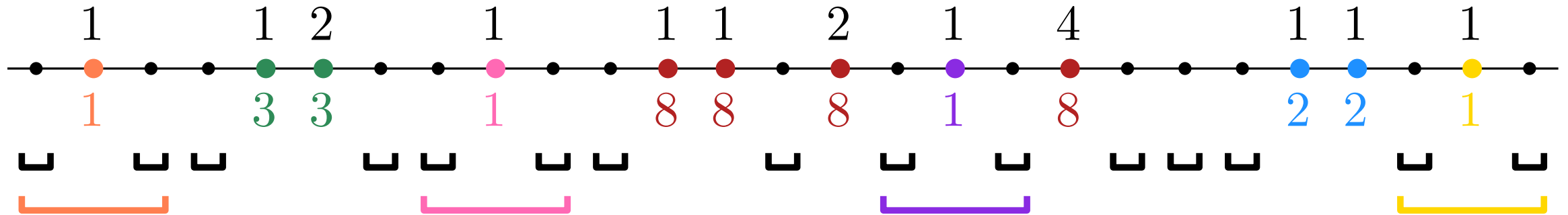
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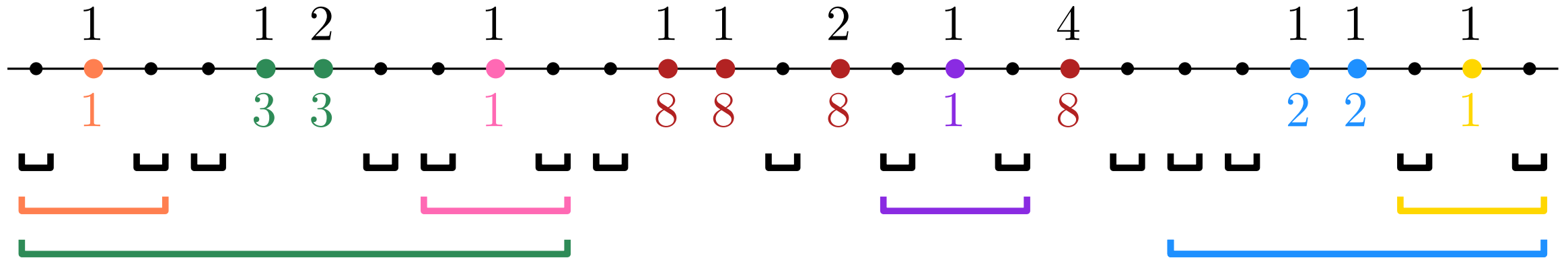




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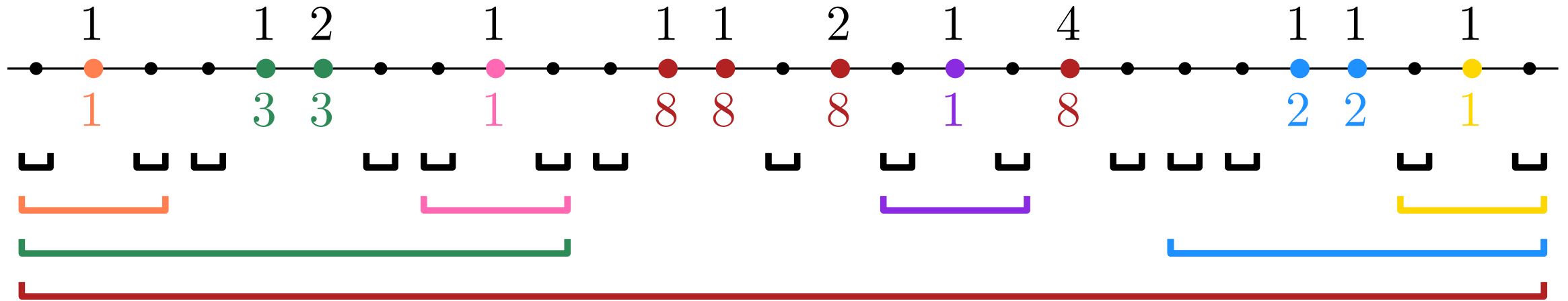
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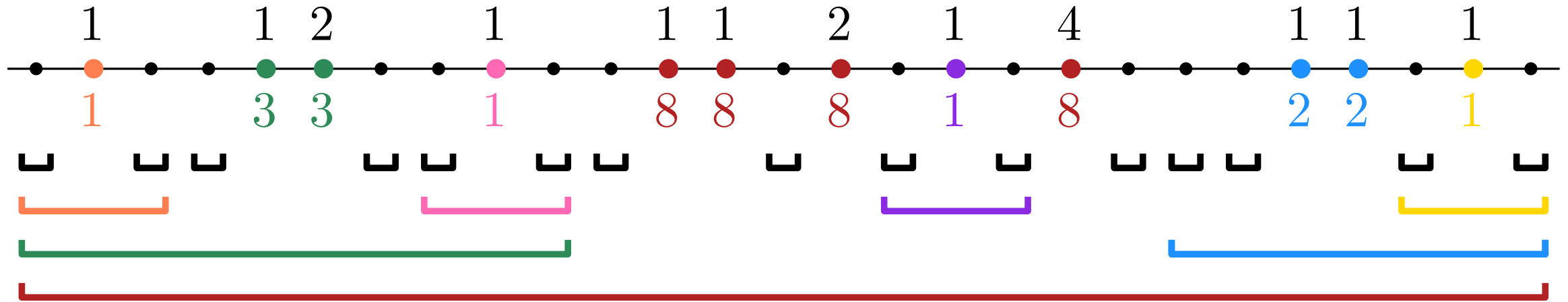
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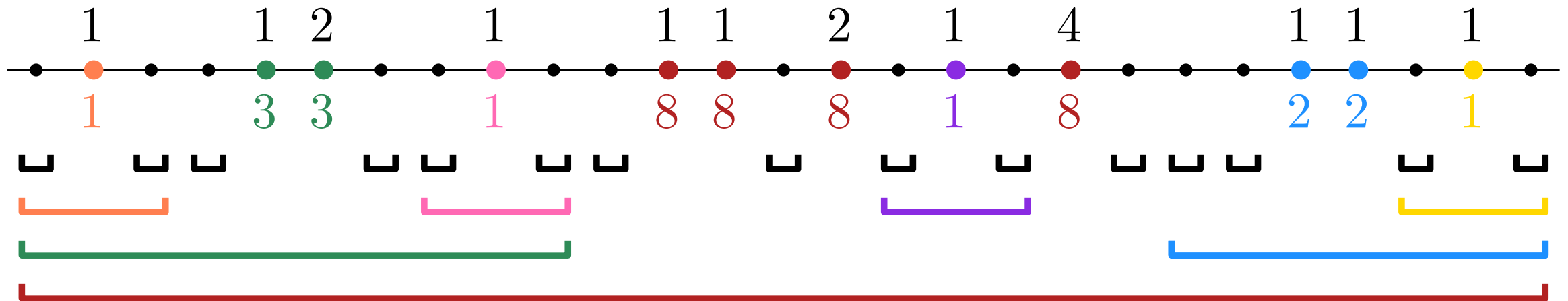


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## Theorem

Suppose  $G$  infinite:

1.  $\forall x \in V : |\mathcal{C}_x| = \infty \Leftrightarrow |\mathcal{S}_x| = \infty \Leftrightarrow \mathcal{S}_x = V$ .
2.  $\mathcal{C}$  has no infinite cluster *iff* there exists an increasing sequence of stable sets  $S_n$  s.t.  $\lim \uparrow S_n = V$ .



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**Proofs:** Multiscale analysis

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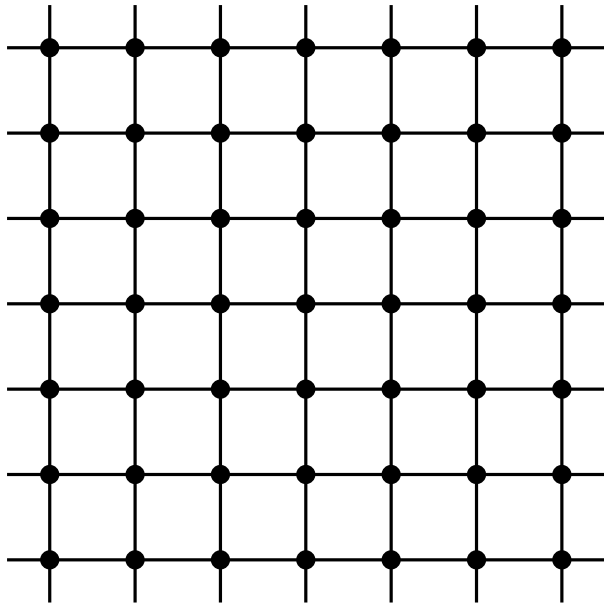
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  1. Binary tree;
  2. Galton-Watson tree with light tails.

# The contact process (Susceptible-Infected-Susceptible)

Epidemic model on graphs introduced by [Harris 74]

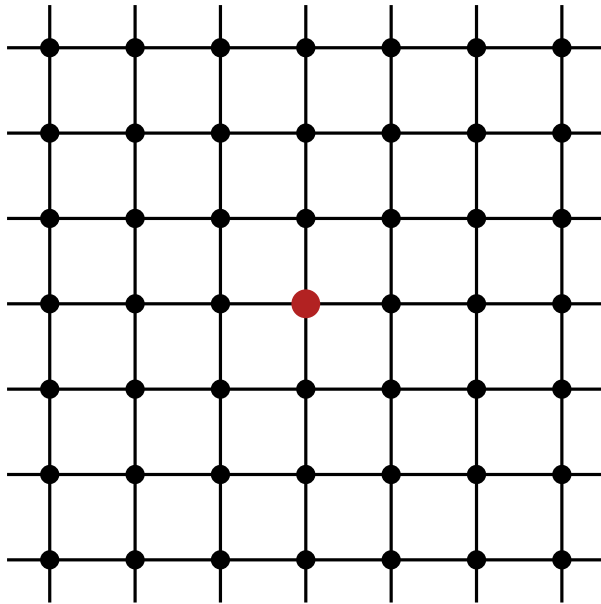


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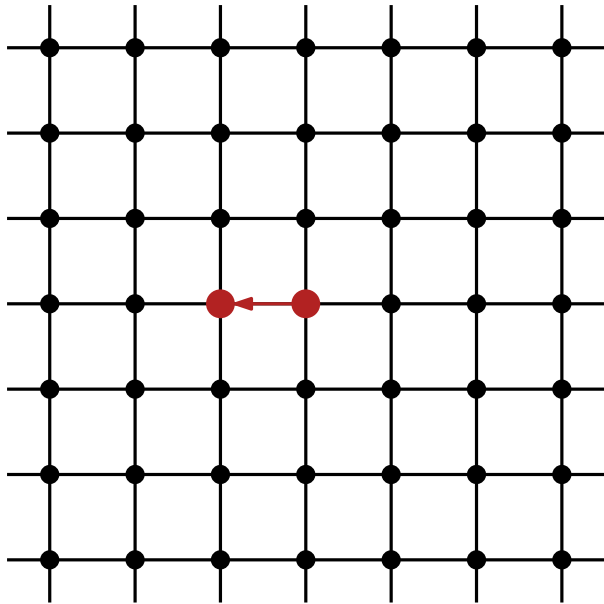


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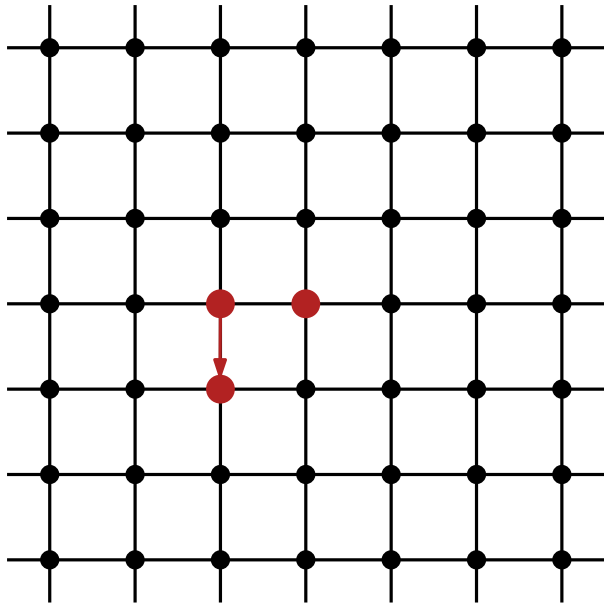


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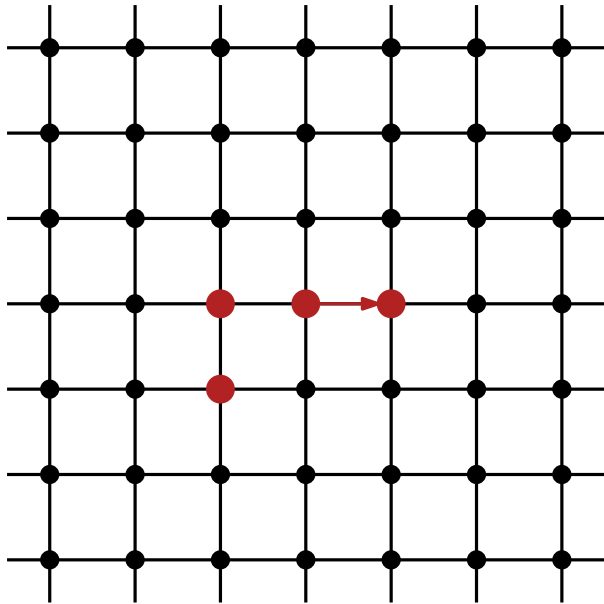


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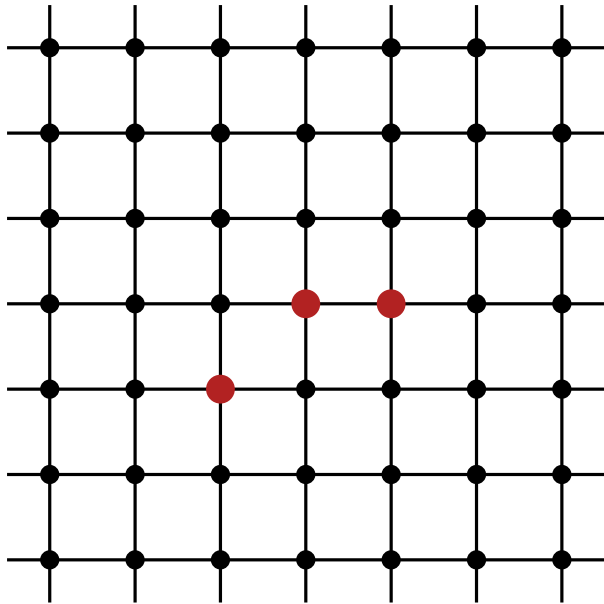


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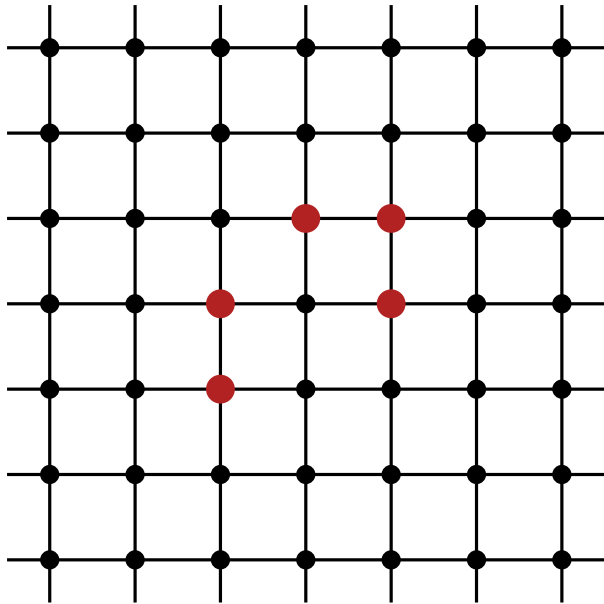
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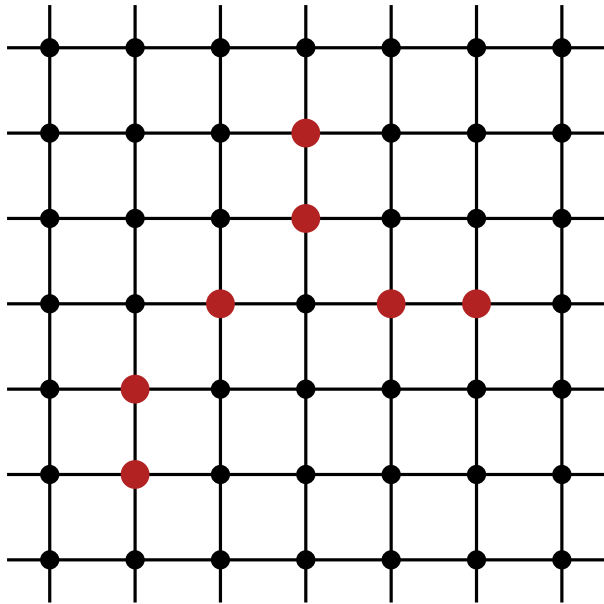


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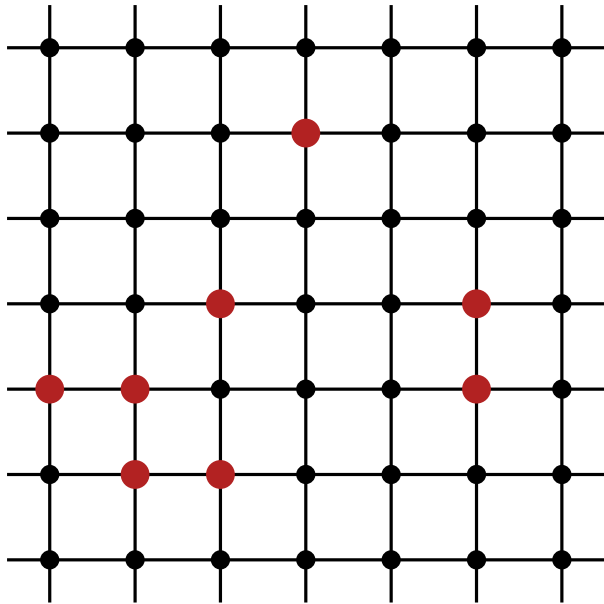


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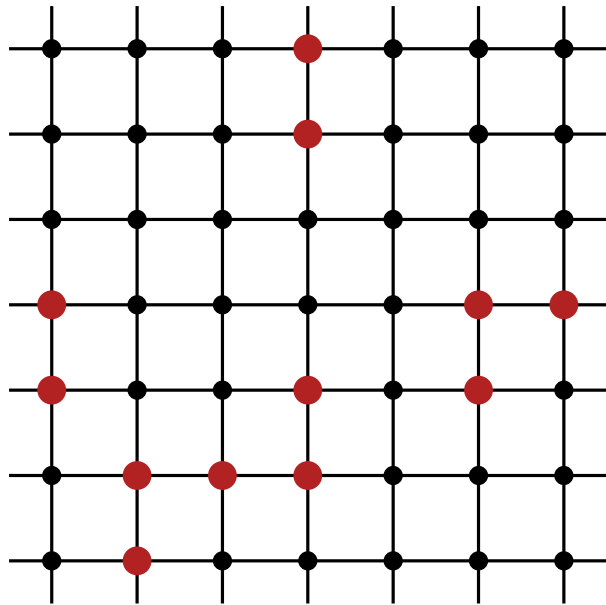
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**Question:** condition on  $G$  to ensure  $\lambda_c > 0$ ?

# The contact process on a graph with bounded degrees

If  $G$  has **bounded** degrees, then  $\lambda_c > 0$ .

Compare with **branching random walk**:

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- No other method to prove that contact process dies out.
- No example of graph with unbounded degrees for which we know  $\lambda_c > 0$ .



# Contact process on geometric graphs

## Theorem

Let  $G$  be either a

- (supercritical) random geometric graph
- Delaunay triangulation

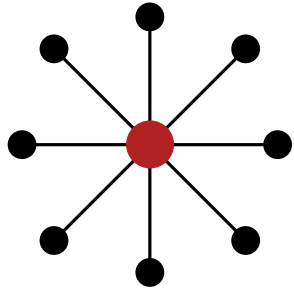
constructed from a Poisson point process on  $\mathbb{R}^d$  with Lebesgue intensity. Then one has  $\lambda_c > 0$ .

## Proof:

Criterion on  $G$  for  $\lambda_c > 0$  in terms of Cumulative Merging Percolation.

# Heuristics for the contact process

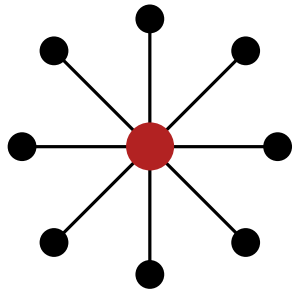
Contact process on a star graph of large degree  $d$ :



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- If  $\lambda > \lambda_c(d)$ , survival time of the process is  $\approx \exp(d)$ .

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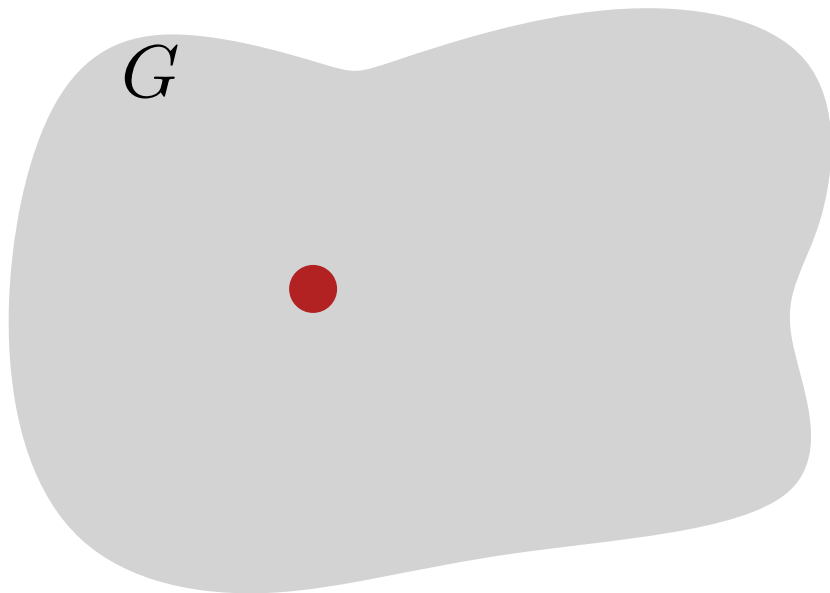
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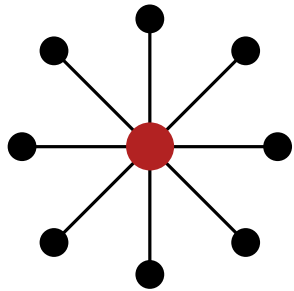
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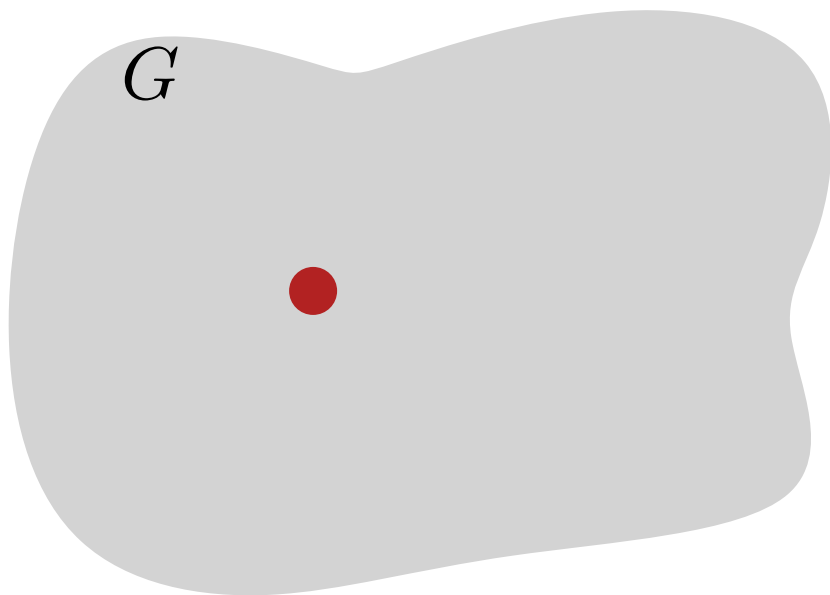
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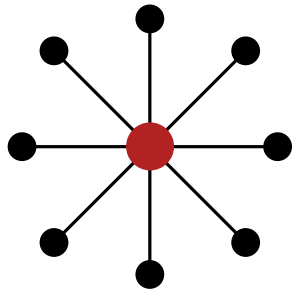
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- Start with only ● infected.
- Force ● to stay infected a time  $\exp(d_0)$ .
- After that time, force the whole star around ● to recover.

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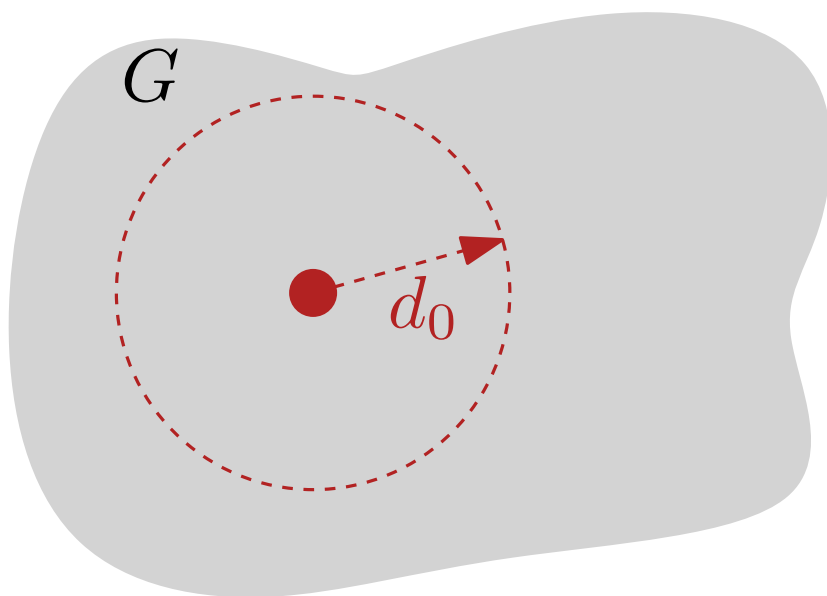
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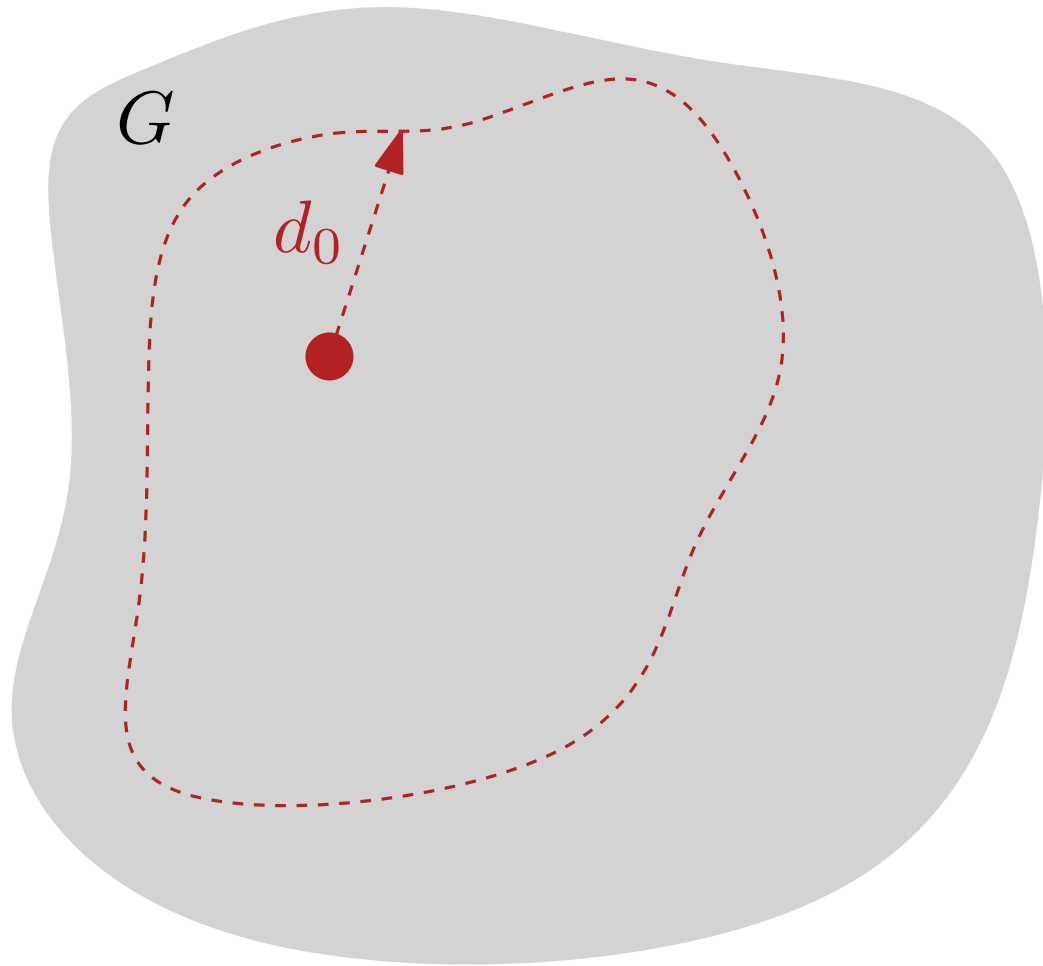


- Start with only  $\bullet$  infected.
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**Maximal distance** reached by infection is  $\approx d_0$ .

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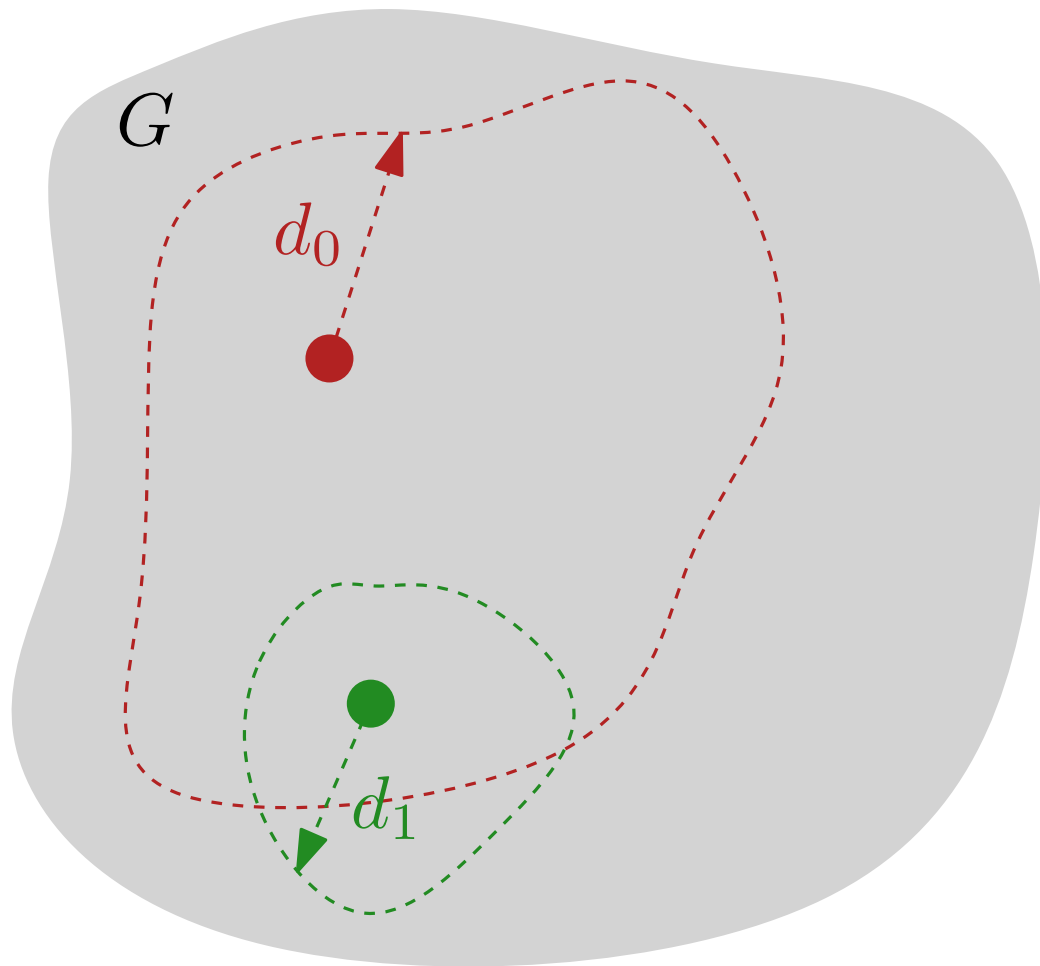


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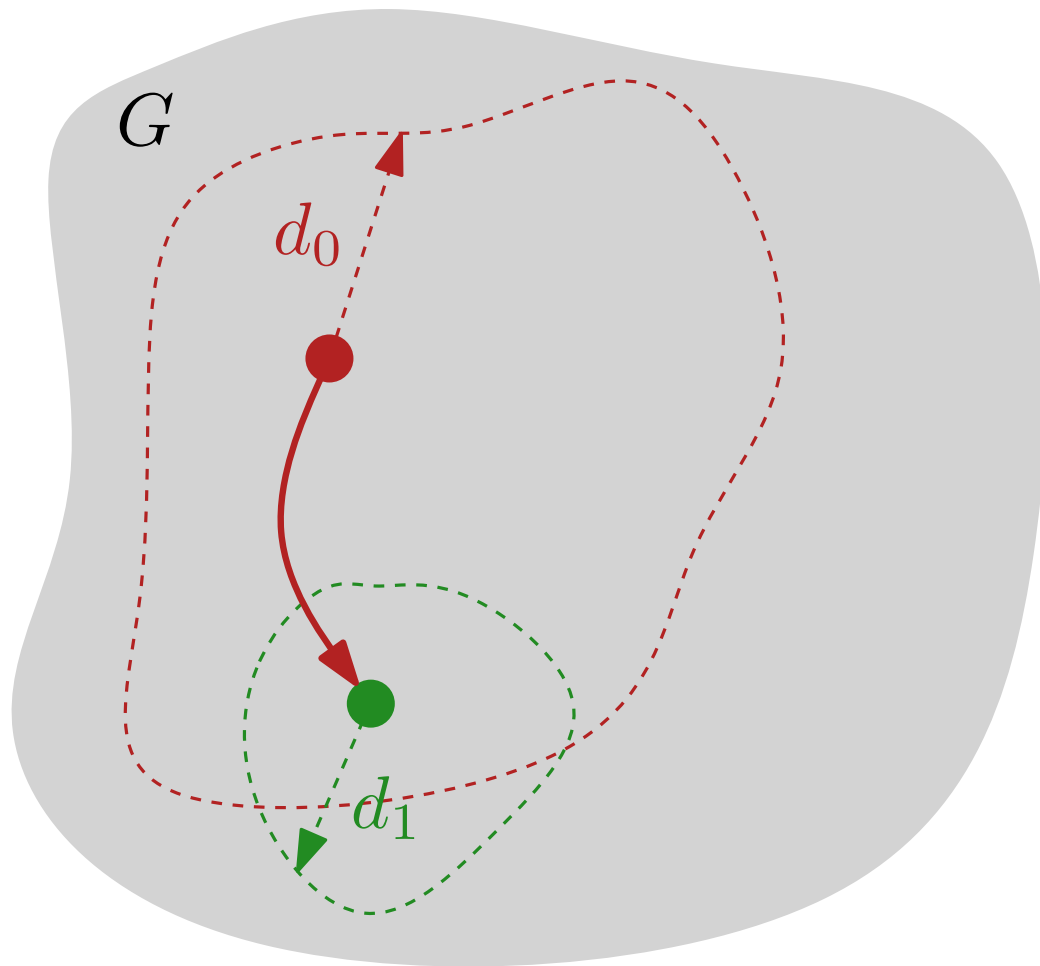


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$\bullet$  cannot send infections to  $\bullet$  and the survival time of the process is  $\approx \exp(d_0) + \exp(d_1) \approx \exp(d_0)$ .

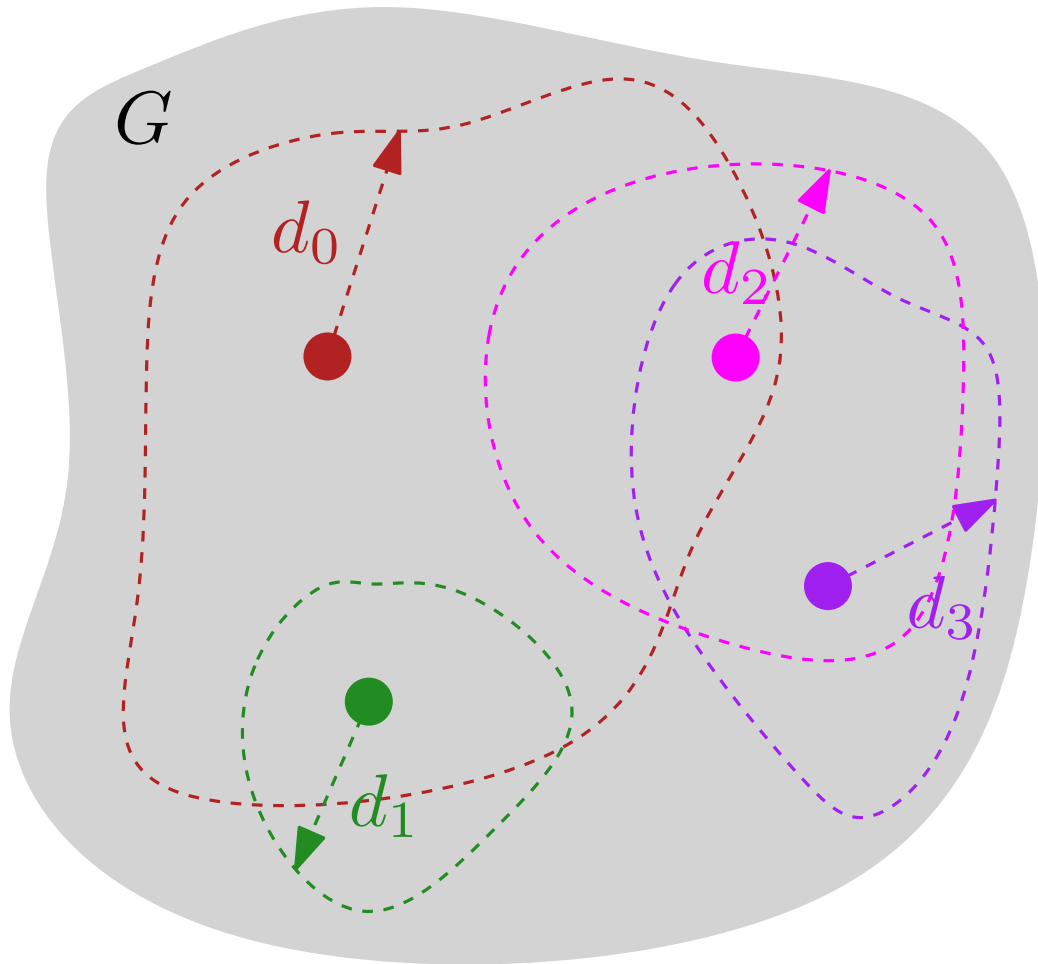


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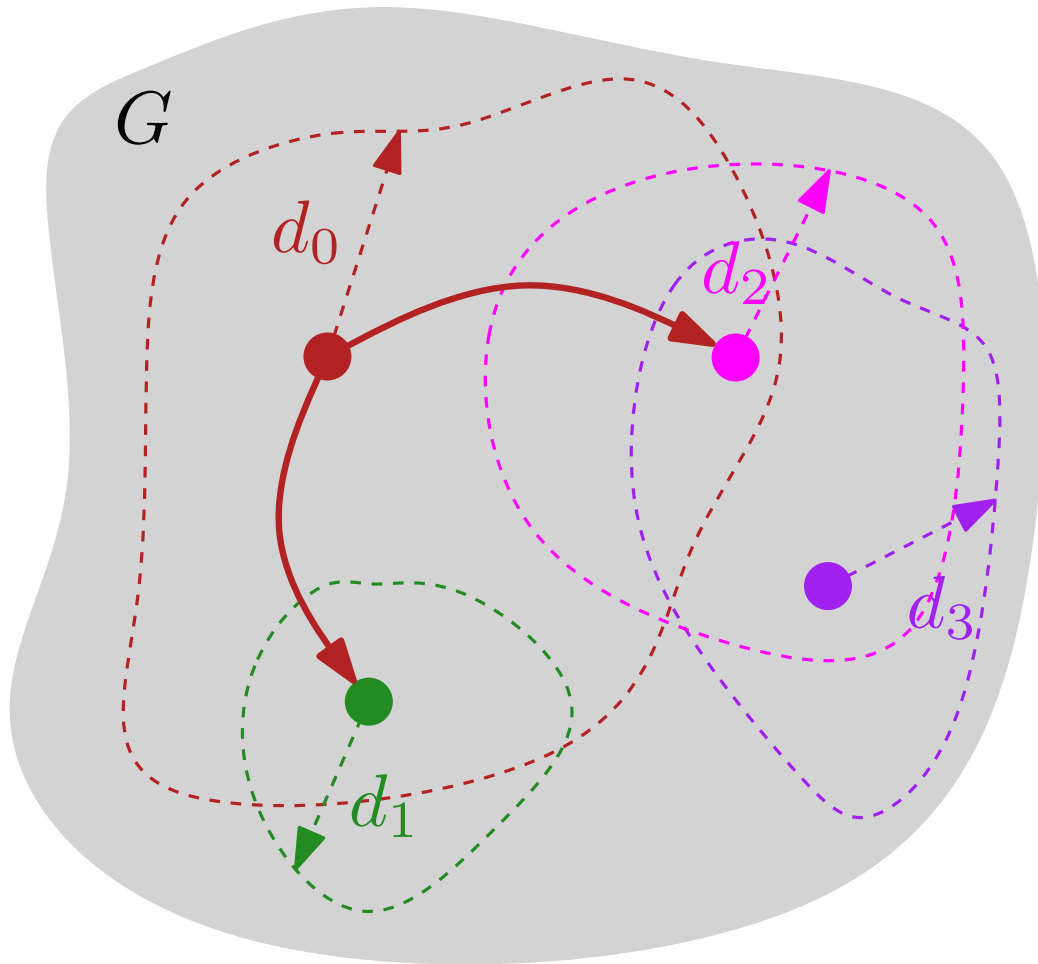


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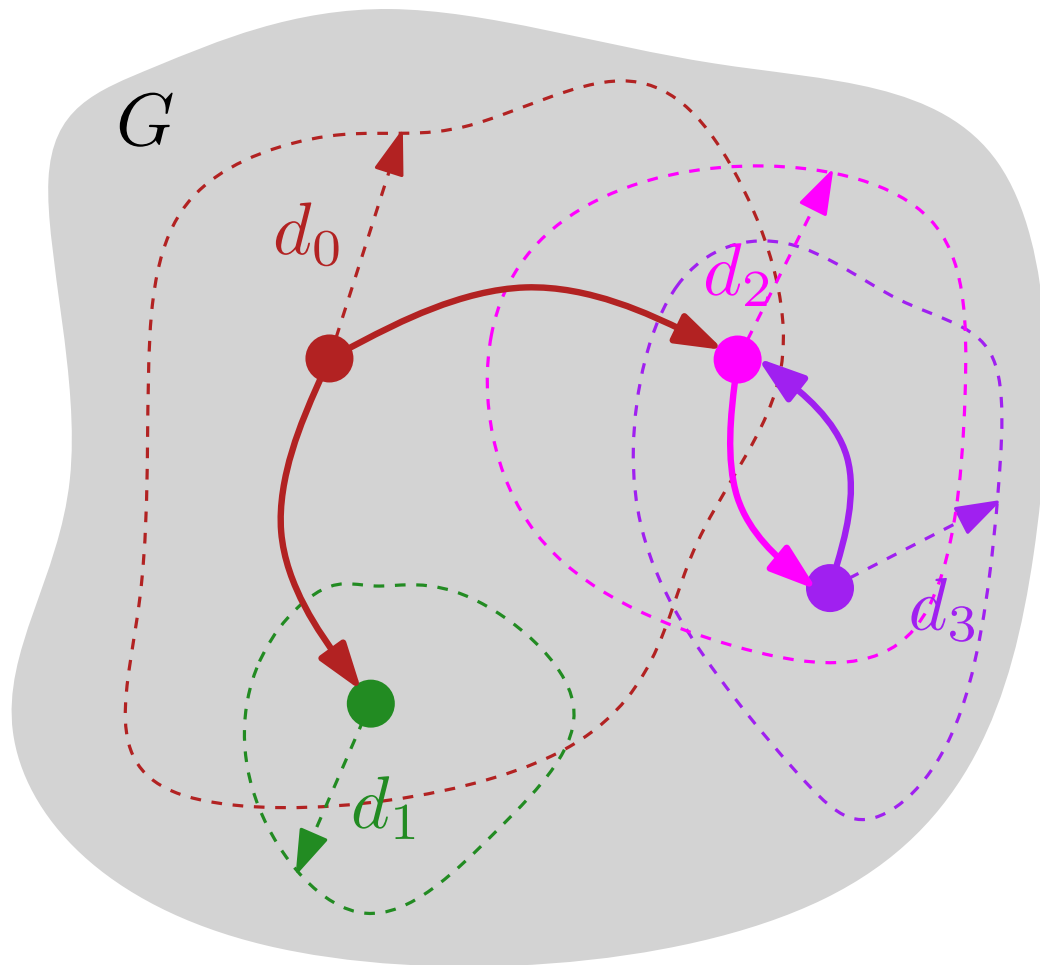
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● and ● interact and their combined survival time is  $\approx \exp(d_2) \times \exp(d_3) = \exp(d_2 + d_3)$



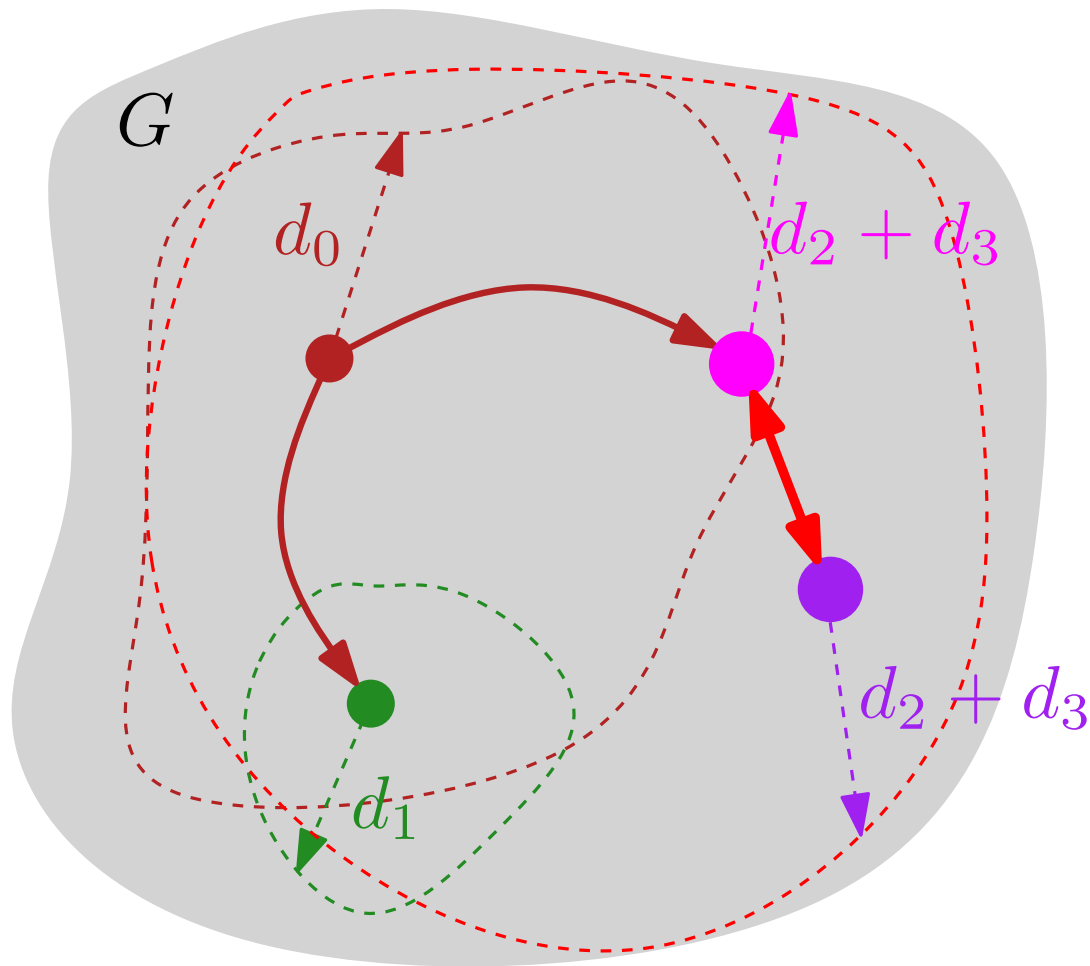
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- $\text{dist}(\text{●}, \text{●}) < d_2 \wedge d_3$

● and ● interact and their combined survival time is  $\approx \exp(d_2) \times \exp(d_3) = \exp(d_2 + d_3)$

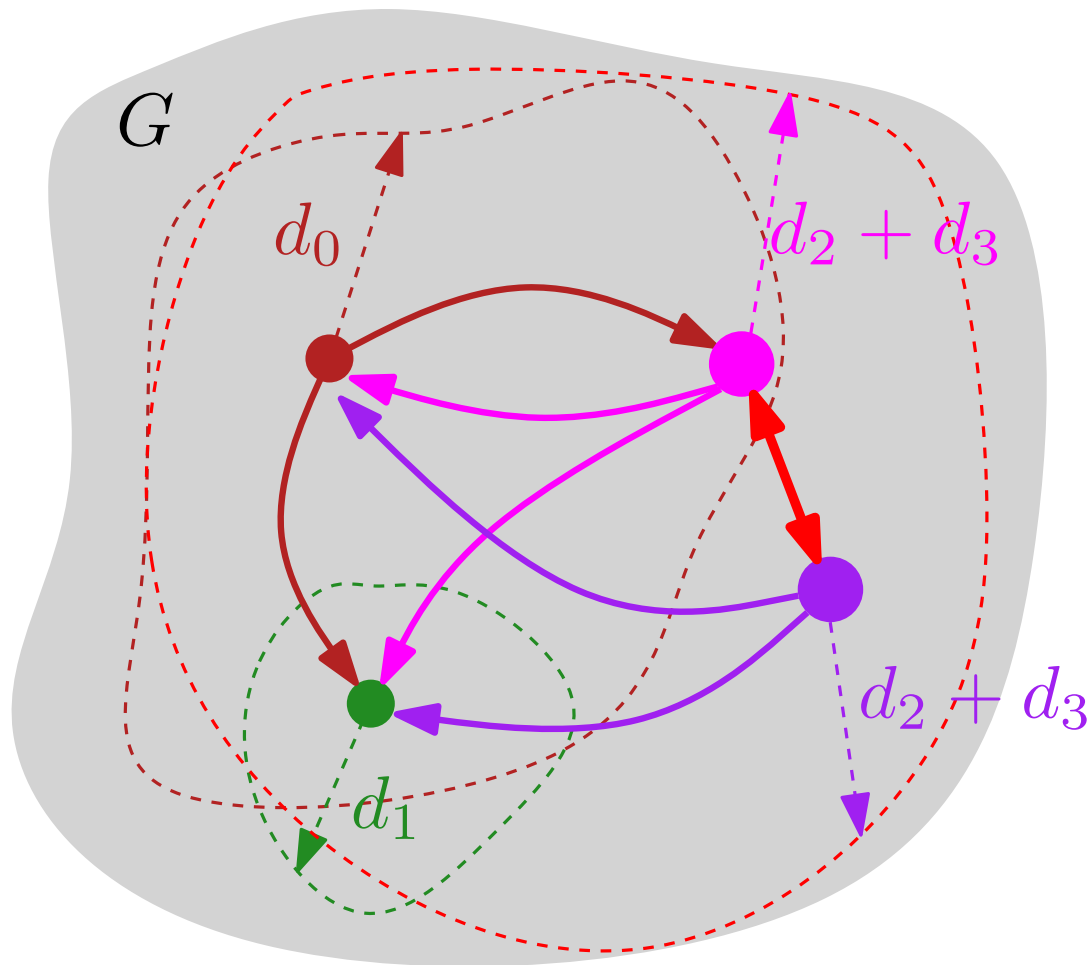


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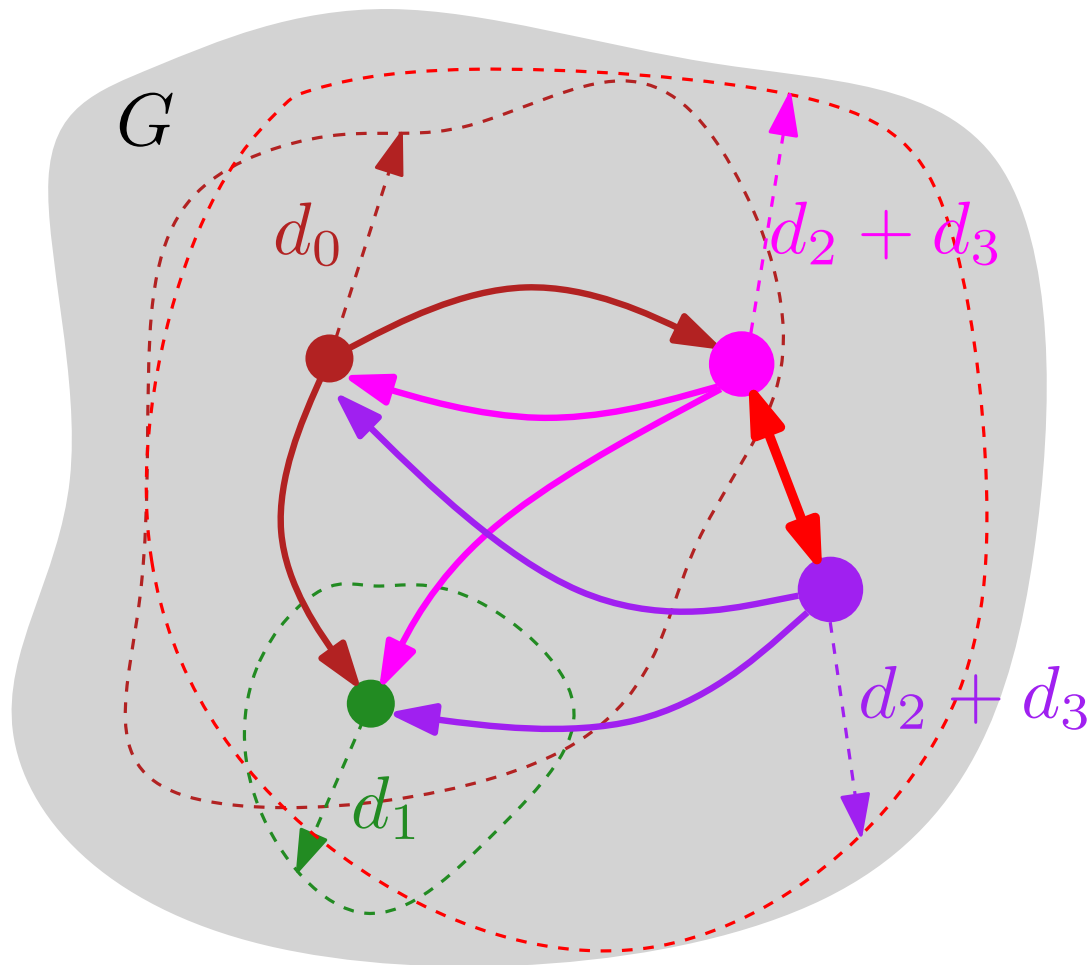
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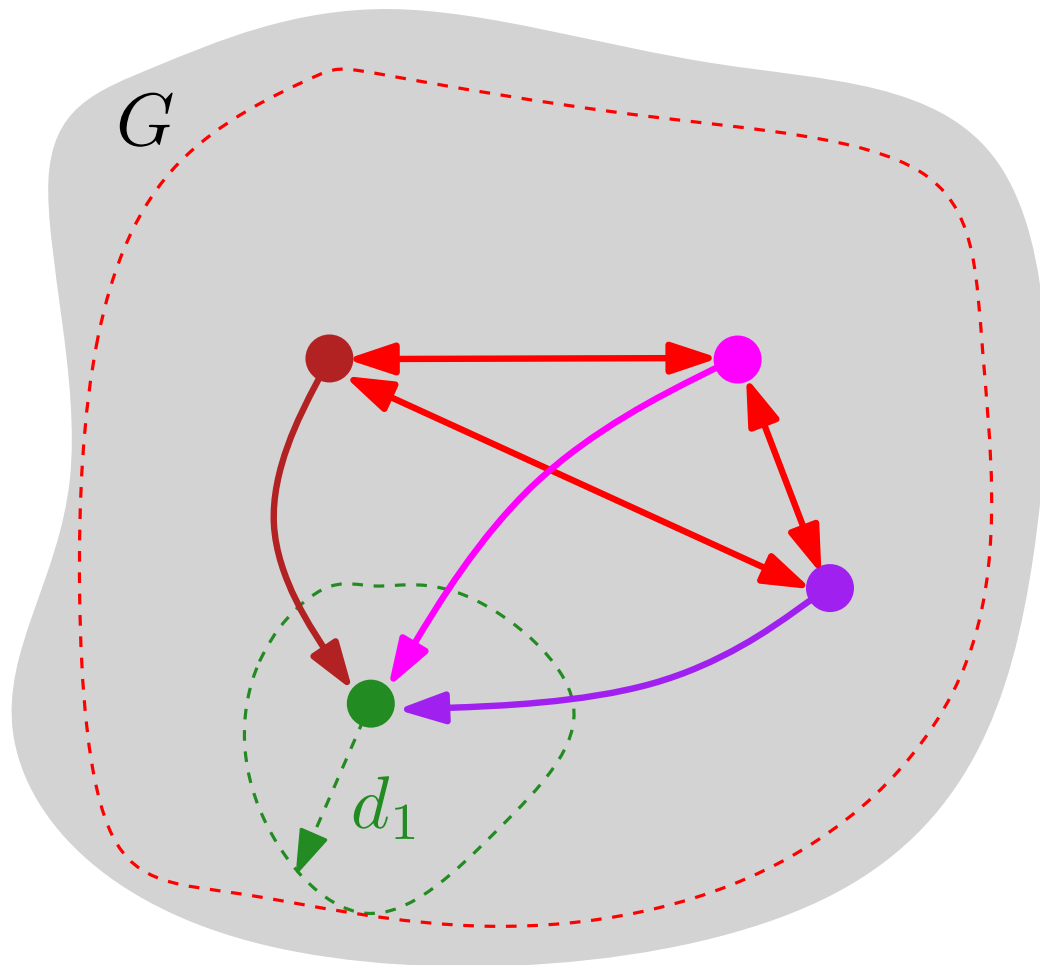
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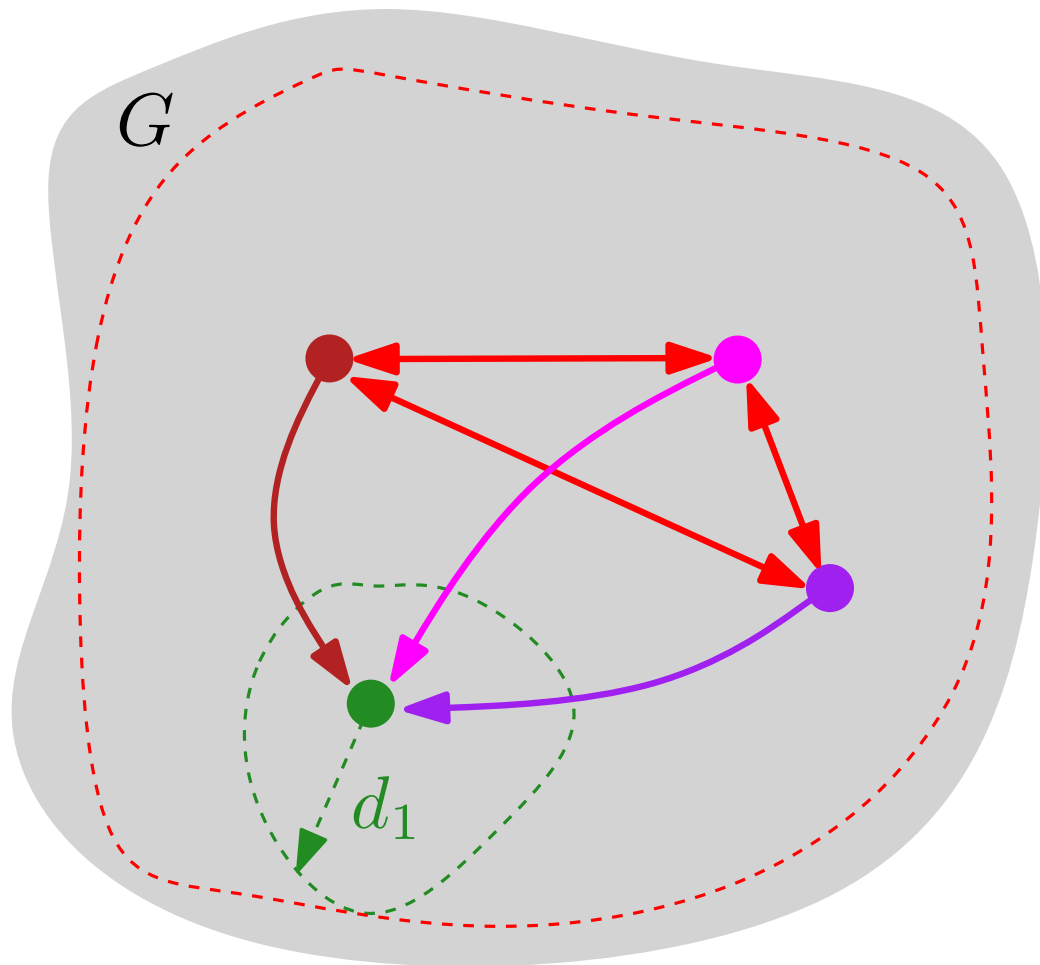
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→ ● still cannot reach the other 3 vertices to interact.



# Cumulative Merging: Link with the contact process

Need to change the definition of admissible partitions:

$\mathcal{P}$  is **admissible** iff  $\forall A, B \in \mathcal{P}$

$$d_G(A, B) > r(A) \wedge r(B) \quad .$$

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## Theorem:

Let  $G = (V, E)$  be **any** locally finite graph. Suppose that, for  $\alpha > 2.5$ , CMP on  $G$  with weights given by:

$$\forall x \in V, r(x) = \begin{cases} \deg(x) & \text{if } \deg(x) > \Delta; \\ 0 & \text{otherwise.} \end{cases}$$

has a non-trivial phase transition (*i.e.*  $\Delta_c < \infty$ ).

Then the contact process on  $G$  has a non trivial phase transition (*i.e.* it dies out for small infection rates).

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**Question:** true for  $\alpha = 1$ ?

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**Thank you for your attention!**