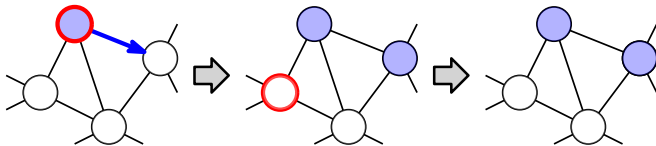


# Evolutionary graph theory

Josef "Pepa" Tkadlec

Mathematical Forum, March 2026



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**Jug lemma:** “On average, an event with probability  $p$  happens after  $1/p$  steps.”

How does stuff propagate  
through networks?

# COVID pandemic



- ▶ Nodes: people. Edges: spending time together
- ▶ Process: people getting Susceptible  $\rightarrow$  Infected  $\rightarrow$  Recovered

## Questions

- ▶ Which region gets hit next? When?
- ▶ Is there enough hospital capacity?
- ▶ Who should we vaccinate first?
- ▶ ...

# How does stuff propagate through networks?



- ▶ epidemiology: coronavirus among humans
- ▶ social dynamics: influence (opinion, gossip, fake news) on social media
- ▶ evolutionary biology: genetic mutations in populations
- ▶ statistical mechanics, percolation, distributed computing, ...

System = a bunch of **nodes** of different **types**.

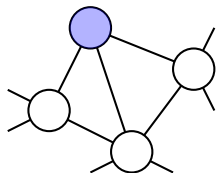
# Overview

- ▶ PUSH protocol for rumor spreading
- ▶ Moran process on graphs

## PUSH protocol for rumor spreading

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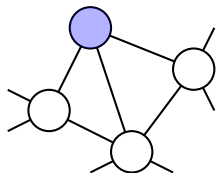
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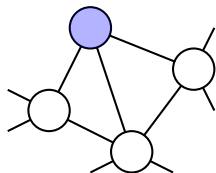
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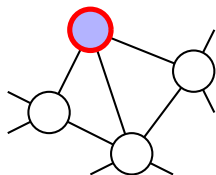
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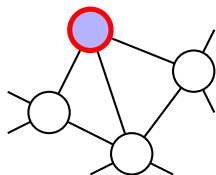
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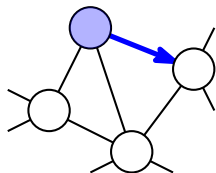
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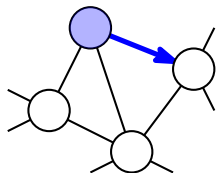
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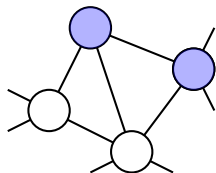
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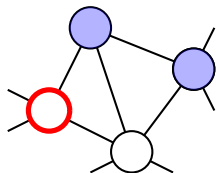
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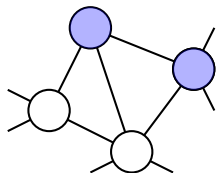
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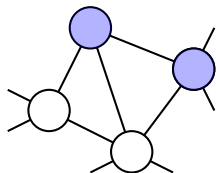
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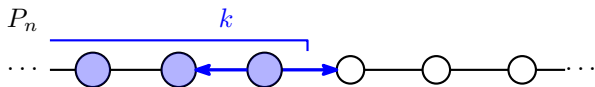
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**Question:** How long until all nodes are informed?

Denote the **expected** number of rounds by  $T(G_n, v)$ .

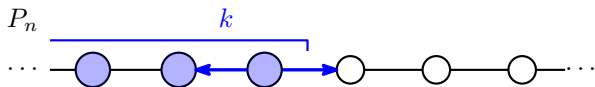
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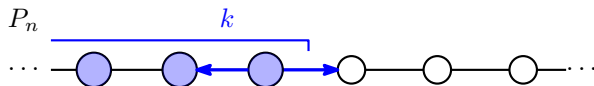
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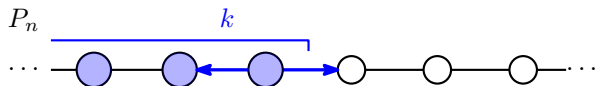
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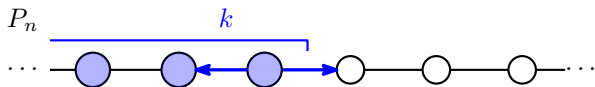


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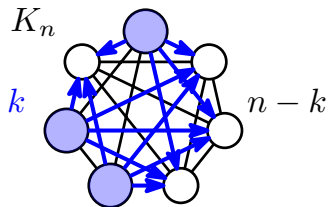


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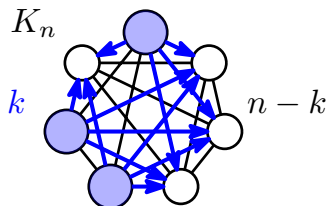
On a **Cycle graph**  $C_n$ :  $T(C_n, v) = \dots = n(n - 1) \approx n^2$ .

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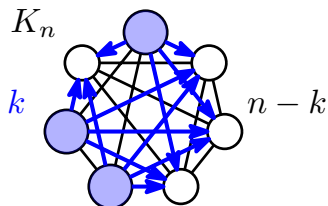
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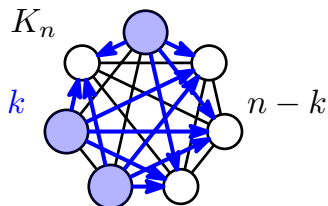
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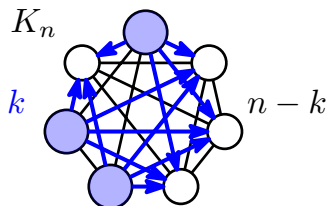


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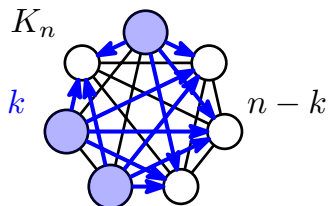
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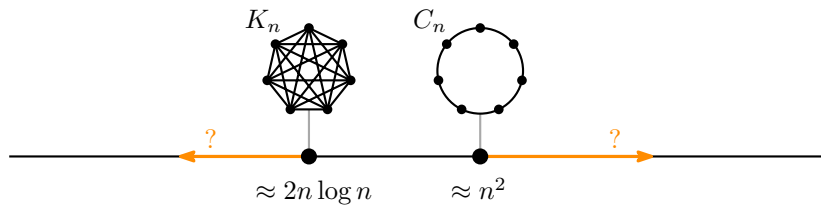
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How fast / slow can a graph be?



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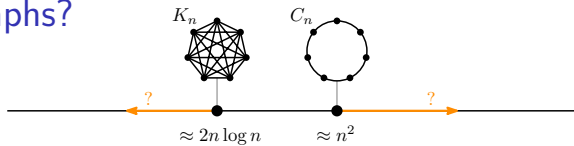
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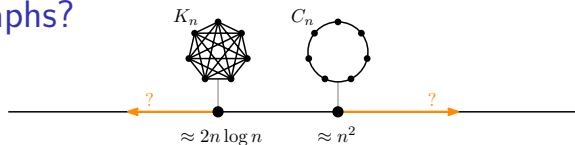
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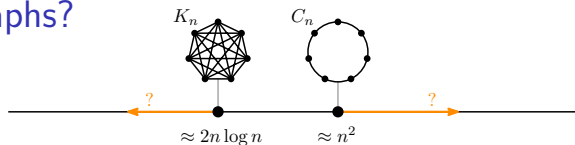


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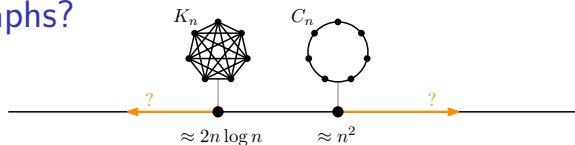
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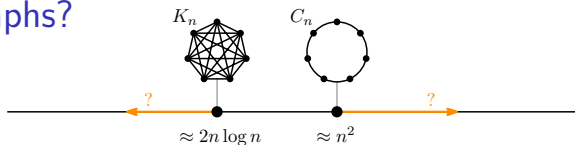


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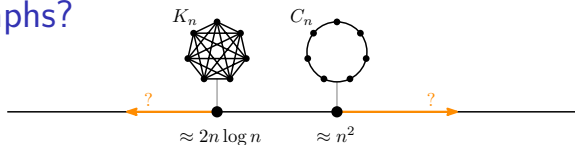
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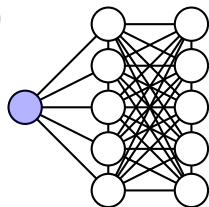
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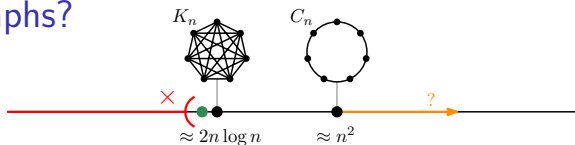
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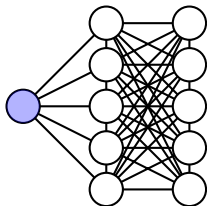
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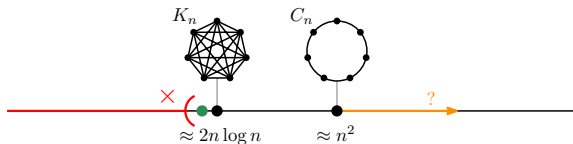
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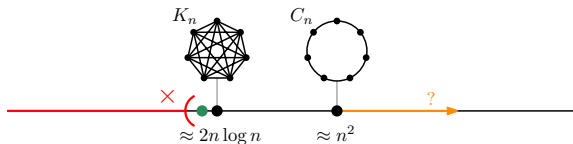


# Slowest graphs?



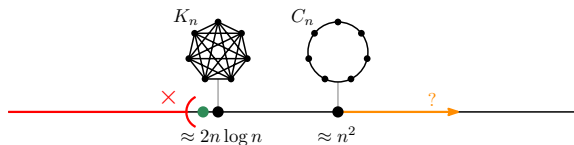
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## Slowest graphs?



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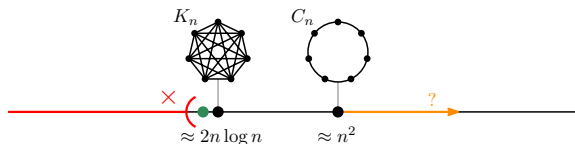
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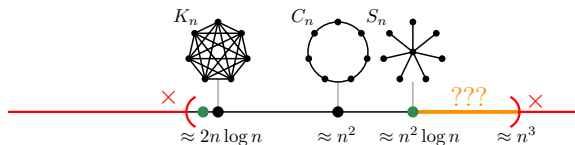


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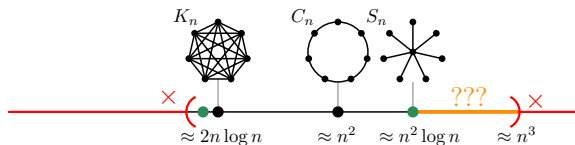


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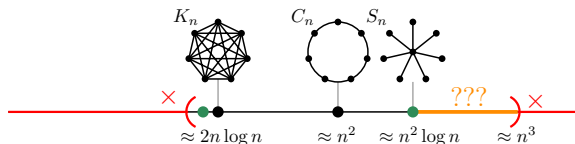
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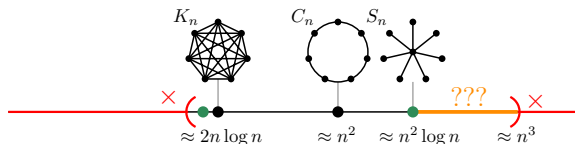
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**Claim [Feige '90].** In fact,  $T(G_n, v) = \mathcal{O}(n^2 \log n)$ .

**Question.** Is  $S_n$  the slowest graph?

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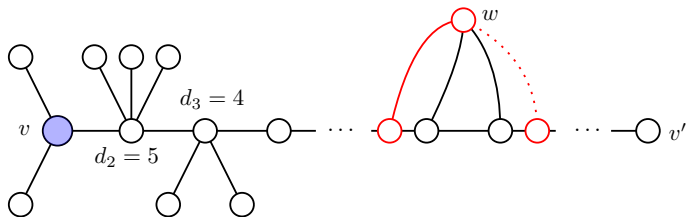
**Question.** Is  $S_n$  the slowest graph? This is **Open**.

## Stronger upper bound

**Key lemma.** For any  $G_n$ ,  $v$ ,  $v'$  we have  $T(v \rightarrow v') \leq 5n^2$ .

**Proof idea:**

- ▶ Take the **shortest** path from  $v$  to  $v'$ .
- ▶ Then  $T(v \rightarrow v') \leq nd_1 + nd_2 + \dots = n \cdot \sum_i d_i$ .



- ▶ Each  $w$  outside the path connects to **at most 3** path nodes!
- ▶ So at most  $3n$  edges connect to the path.
- ▶ So  $\sum_i d_i \leq 5n$  and  $T(v \rightarrow v') \leq n \cdot 5n = 5n^2$ .

## Stronger upper bound, cont'd

Recall:

**Key Lemma.** For any  $G_n$ ,  $v$ ,  $v'$  we have  $T(v \rightarrow v') \leq 5n^2$ .

**Claim [Feige '90].** We have  $T(G_n, v) = \mathcal{O}(n^2 \log n)$ .

**Proof idea:**

- ▶ Split into **phases** of  $10n^2$  steps each.
- ▶ Fix  $v'$ . By **Lemma** and **Markov inequality**,  $v'$  is informed with probability  $\geq 1/2$ .
- ▶ So after  $k$  phases,  $v'$  is informed unless  $1/2^k$ .

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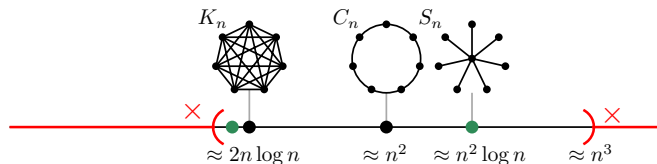
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- ▶ By **Union bound**,  $k$  rounds suffice unless  $n/2^k$ , so

$$\begin{aligned} T(G_n, v) &\leq 10n^2 \cdot \log_2 n + 10n^2 \cdot \sum_{k \geq \log_2 n} \frac{n}{2^k} \\ &= 10n^2 \log_2 n + \left(1 + \frac{1}{2} + \frac{1}{4} + \frac{1}{8} + \dots\right) = 20n^2 \log_2 n. \end{aligned}$$

# Landscape



## Open Questions:

1. What is the fastest graph?
2. What is the slowest graph? Is it the Star  $S_n$ ?
3. Efficient algorithm to compute  $T(G_n, v)$ , for any  $G_n$ ?

# Overview

- ▶ PUSH protocol for rumor spreading
- ▶ Moran process on graphs

# Overview

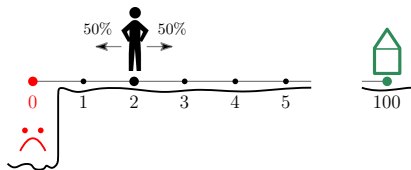
- ▶ PUSH protocol for rumor spreading
  - ▶ “progressive”: only 1 color spreads
  
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# Overview

- ▶ PUSH protocol for rumor spreading
  - ▶ “progressive”: only 1 color spreads
  
- ▶ Moran process on graphs
  - ▶ both colors fight!

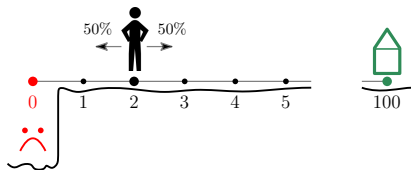
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**Puzzle.** A drunk person, repeatedly stepping left/right 50 : 50.  
How likely are they to make it home?



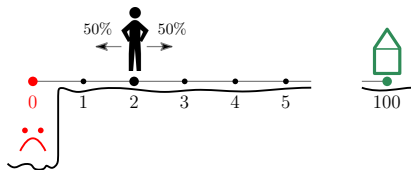
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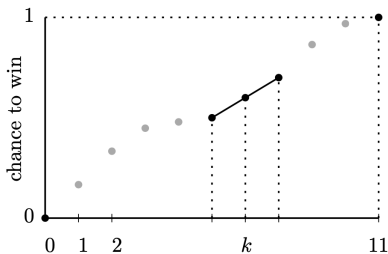
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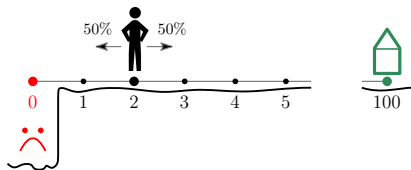
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Then  $p_k = \frac{1}{2}p_{k-1} + \frac{1}{2}p_{k+1}$ .



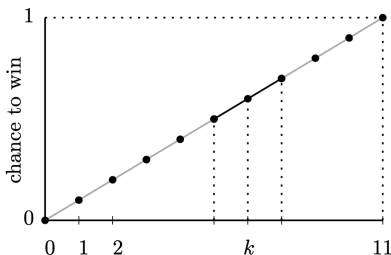
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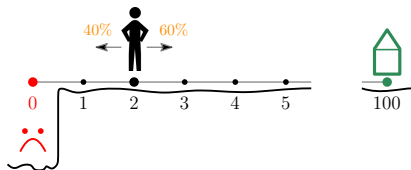
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## Puzzle: Biased Random Walk

Puzzle. Not completely drunk, steps left/right 40 : 60.

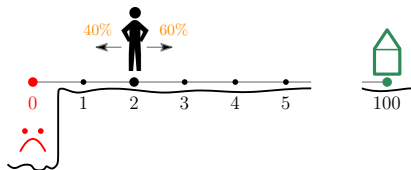
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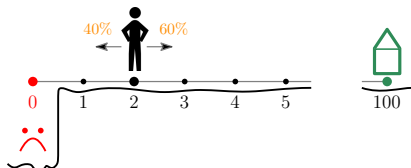
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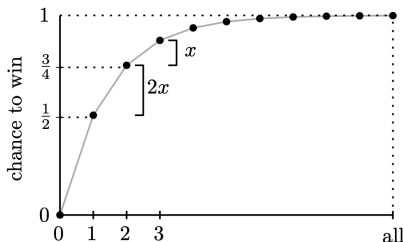
**Puzzle.** Not completely drunk, steps left/right 40 : 60.

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**Proof idea:** This time  $p_k = 0.4 \cdot p_{k-1} + 0.6 \cdot p_{k+1}$ .

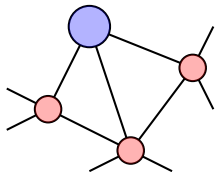
If it were  $\frac{1}{3} : \frac{2}{3}$ , the answer would be  $\approx 75\%$





## Model: Moran process on a graph

[Moran '58] [LHN, '05] A graph  $G_n = (V, E)$  on  $n$  nodes.

- ▶ Nodes: individuals (fitness: residents 1, mutants  $r \geq 1$ )

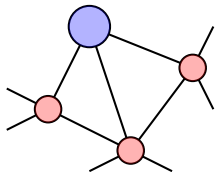




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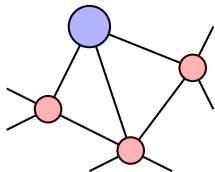




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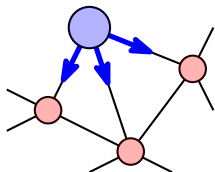




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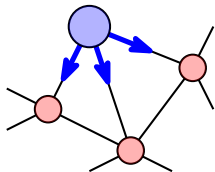




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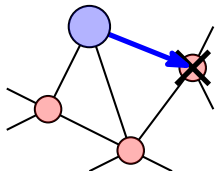




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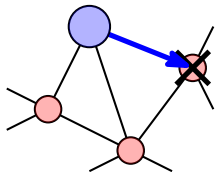




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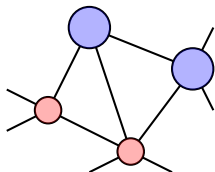




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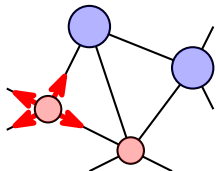




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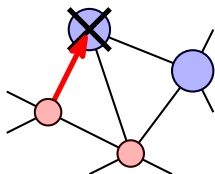




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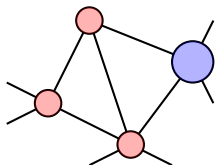




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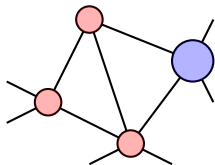
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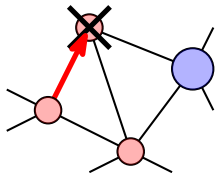
	fitness
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1. It is stochastic (random).
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4. Eventually, all nodes become the same type (no mutation).

Quantities of interest:

1. **Fixation probability**  $fp^r(G_n, v)$ : How likely is it that blue wins?
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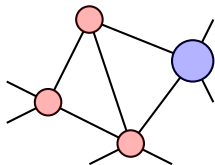
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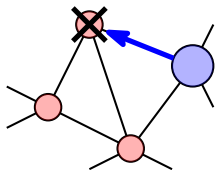
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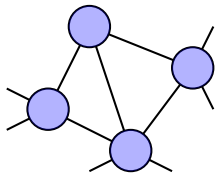
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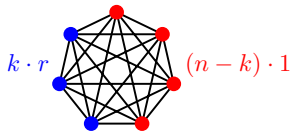
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## Moran process on $K_n$ ( $r > 1$ )

Idea: Track #mutants and instead of  $p_k$ , have  $p_k^+$  and  $p_k^-$ .

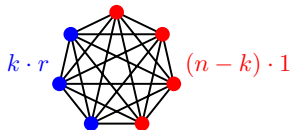


$$F = kr + (n - k)$$

$$\left. \begin{aligned} p_k^+ &= \frac{kr}{F} \cdot \frac{n-k}{n-1} \\ p_k^- &= \frac{n-k}{F} \cdot \frac{k}{n-1} \end{aligned} \right\} \frac{p_k^+}{p_k^-} = r$$

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- ▶ In each step, we are  $r$ -times more likely to gain than to lose a mutant.
- ▶ So like the drunk with a sense of direction.

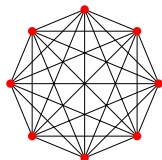


**Claim.**  $\text{fp}^r(K_n) = \frac{1-1/r}{1-1/r^n} \rightarrow_{n \rightarrow \infty} \boxed{1 - 1/r}$ .

## Regime $r > 1$ fixed, $n$ large

1. Baseline: Complete graph  $K_n$ . We have  $\text{fp}^r(K_n) \rightarrow 1 - 1/r$ .

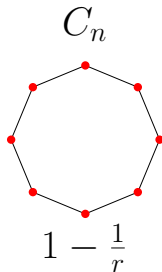
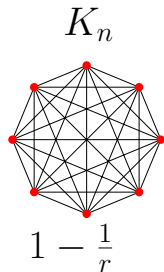
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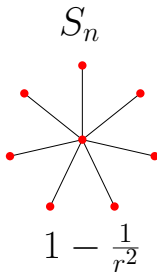
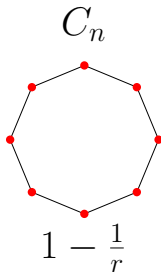
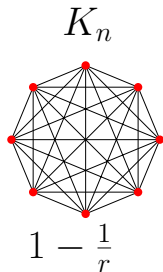
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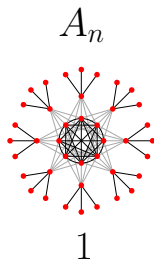
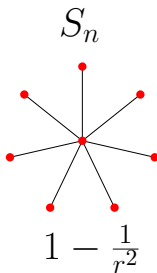
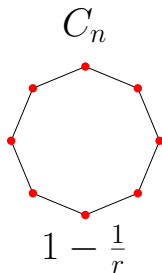
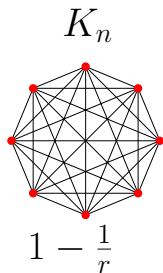
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4. There exist **strong amplifiers**: graphs  $A_n$  with  $\text{fp}^r(A_n) \rightarrow_{n \rightarrow \infty} 1$ , for any  $r > 1$ .



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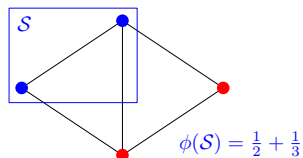
**Answer:**  $(B - \phi_0)/\delta \dots$  under mild assumptions.

## FPRAS for undirected graphs

Theorem [DGMRSS, '14]. Fix  $r \geq 1$ . Then  $T^r(G_n, r) = \mathcal{O}(n^4)$ .

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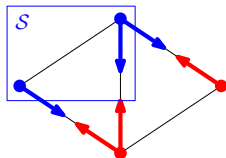


Proof idea:

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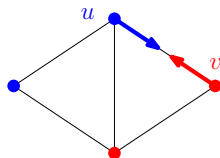


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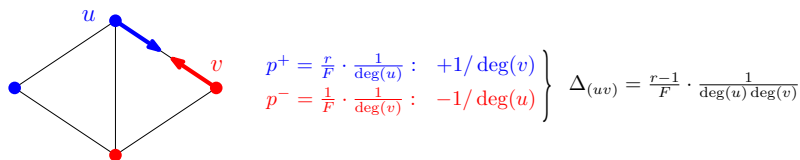


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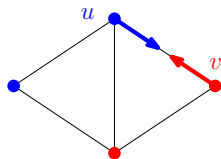


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$$\left. \begin{aligned} p^+ &= \frac{r}{F} \cdot \frac{1}{\deg(u)} : +1/\deg(v) \\ p^- &= \frac{1}{F} \cdot \frac{1}{\deg(v)} : -1/\deg(u) \end{aligned} \right\} \Delta_{(uv)} = \frac{r-1}{F} \cdot \frac{1}{\deg(u)\deg(v)}$$

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2. Claim: contribution of any fixed “active” edge  $(u, v)$  is  $+\Omega(1/n^3)$ .
3. But  $\phi(\mathcal{S}) \in [0, n]$ , so  $\phi$  “can’t keep increasing for too long”.
4.  $\Rightarrow T^r(G_n) = \mathcal{O}(n^4)$ .

## Summary

- ▶ Fixation probability can be approximated by simulations.
- ▶ Fixation time is between  $\Omega(n \log n)$  and  $\mathcal{O}(n^{3+\epsilon})$  [GLR '19].

### Open questions:

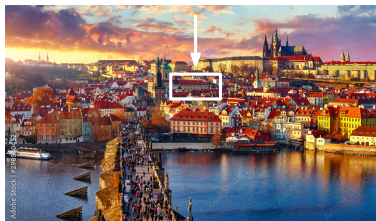
1. Computing  $\text{fp}^r(G_n, v)$  exactly?
2. Approximations on directed graphs?
3. Adversarial amplifier?
4. Location-dependent fitness?
5. #mutants at time  $T$ ?
6. More colors?
7. ...

# Wrap-up

- ▶ How does stuff spread through networks?
- ▶ Even for the simplest model, basic questions are open.
- ▶ **Puzzles:** Jug lemma, Coupon collector, (biased) Random walk



EGT-Prague



Computer Science Institute, Charles Uni

# Thank you

