# Expanders via Local Edge Flips

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#### **Outline**

- How can we construct an expander locally?
  Problem motivation and related works
- A simple distributed protocol The switch and the flip protocols
- A new analysis for the two protocols
  Obstacles in the analysis and new approach for the problem
- Conclusions and future directions
  Open problems

# How can we construct an expander locally?

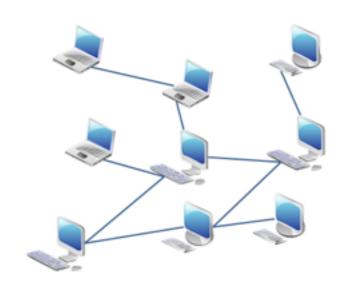
## Why is it interesting?

#### Distributed system

P2P networks

Sensor networks

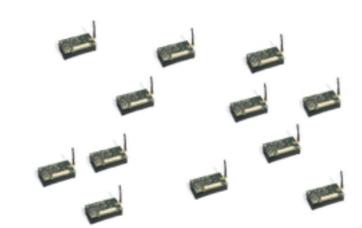
Asynchronous system



#### Benefits

Efficient

Robust



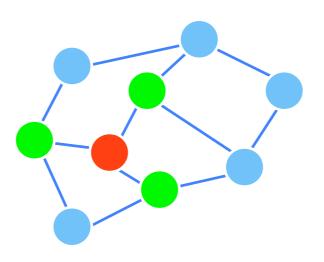
#### New challenges

Important to construct quickly good network structure
Only local communication

## Local graph algorithms

#### Local algorithms

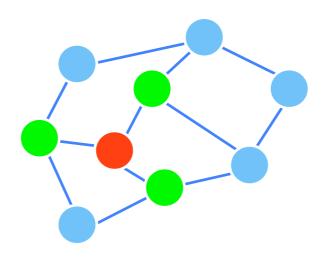
Algorithms based on *local* message passing among nodes



## Local graph algorithms

#### Local algorithms

Algorithms based on *local* message passing among nodes



#### Advantages

Applicable to large scale graphs

Fast, easy to implement in parallel (MapReduce, Hadoop, Pregel...)

### **Problem**

Starting from any connected graph is it possible to construct an expander locally?

#### **Previous work**

#### SKIP+: A Self-Stabilizing Skip Graph.

R. Jacob, A. W. Richa, C. Scheideler, S. Schmid and H. Täubig.

**J. ACM** 61(6): 36:1-36:26 (2014)

In the Local model it is possible to build an expander locally in  $O(\log^2 n)$ 

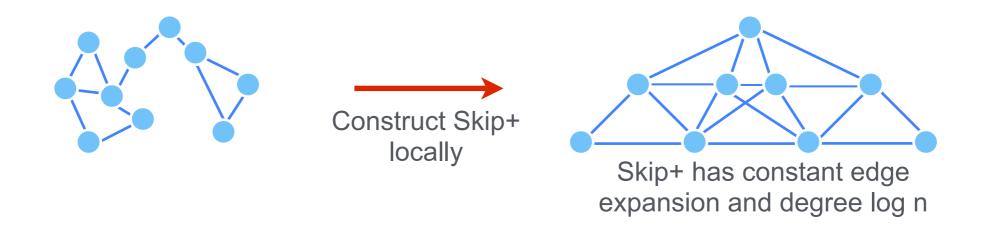
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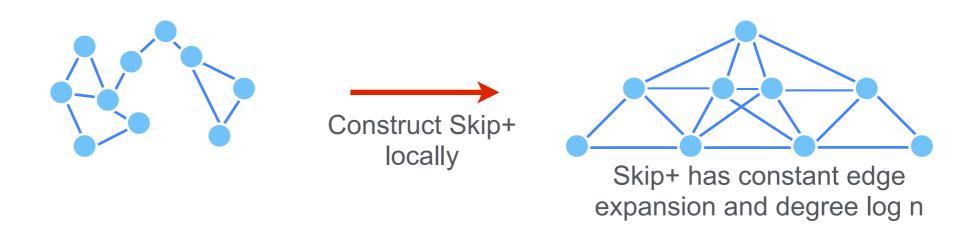
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#### Limitations:

- Using this technique it is not possible to obtain an algebraic expander
- In any round nodes can exchange arbitrary large messages
- Memory needed by a single node in any round is not bounded
- Synchronous model, complex algorithm

#### **Problem**

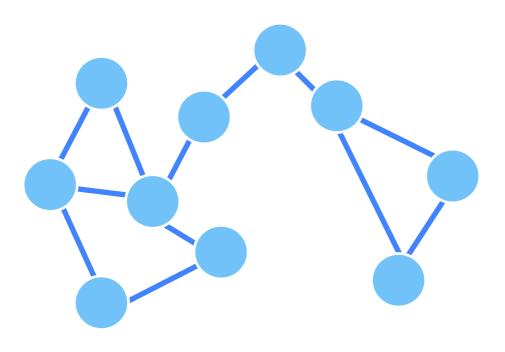
Starting from any connected graph is it possible to define a simple rule to construct an expander locally?

# A simple distributed protocol

[McKay, Congressus Numerantium 1981]

A simple protocol:

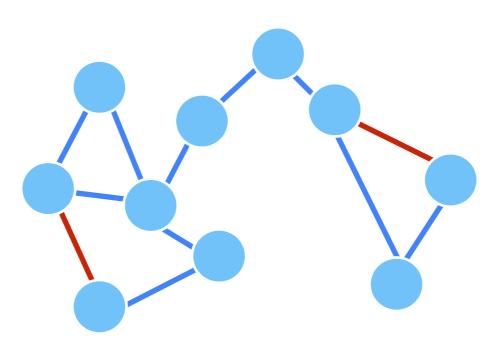
Pick two edges at random and invert their endpoints



[McKay, Congressus Numerantium 1981]

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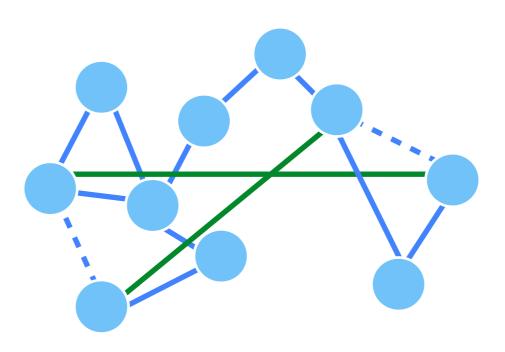
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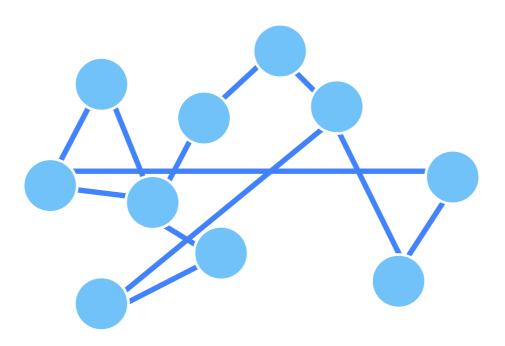
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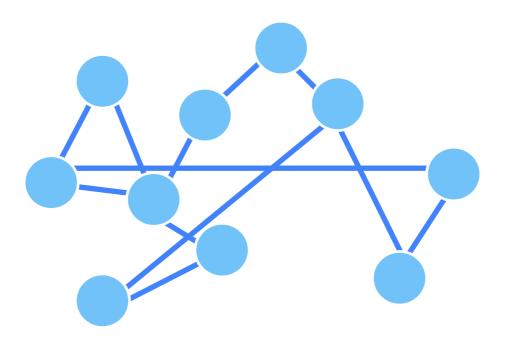
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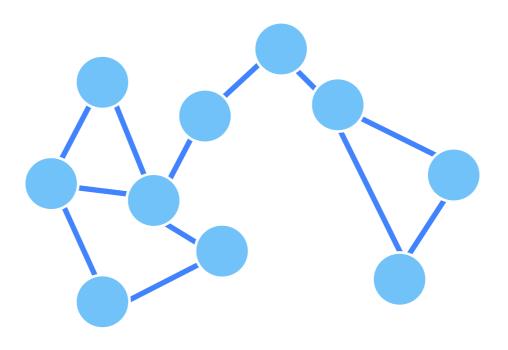
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Limitation
It is not local
It may disconnect the graph

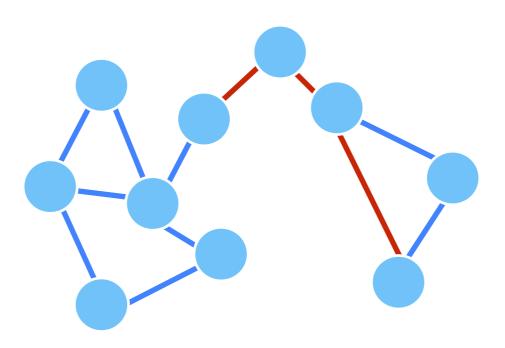
[Mahlmann and Schindelhauer, SPAA 2005]

Pick a random length 3 path and invert its endpoints



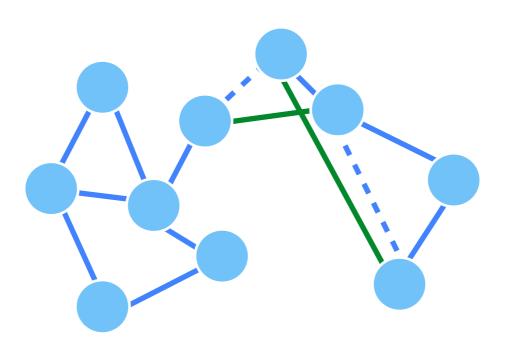
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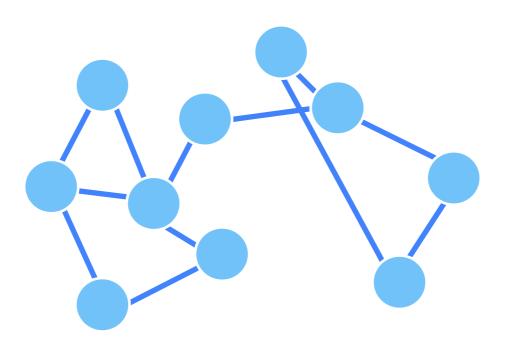
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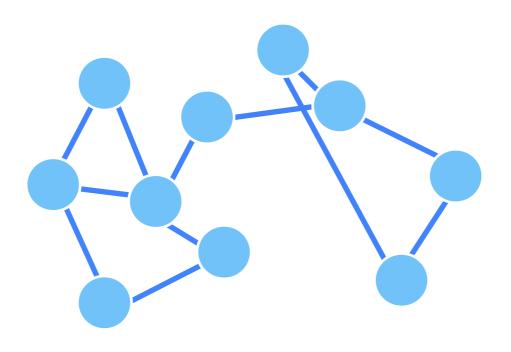
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Experimentally it seems to be really fast

### What is known about them?

[Cooper, Dyer and Greenhill, SODA 2005]

For d-regular graph the switch protocol converges to the configuration model in  $\tilde{O}\left(n^8d^{15}\right)$  steps.

[Greenhill, SODA 2015]

For non regular graph with max degree in  $O\left(\sqrt{m}\right)$  the switch protocol converges to the configuration model in  $\tilde{O}\left(m^{10}d_{max}^{14}\right)$  steps.

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[Mahlmann and Schindelhauer, SPAA 2005]

For d-regular graph the flip protocol converges to the configuration model.

[Feder, Guetz, Mihail, and Saberi, FOCS 2006]

For d-regular graph the flip protocol converges to the configuration model in  $\tilde{O}\left(d^{34}n^{36}\right)$  steps.

[Cooper and Dyer, PODC 2009]

For d-regular graph the flip protocol converges to the configuration model in  $\tilde{O}(d^{23}n^{17})$  steps.

## How do they perform in practice?

[Mahlmann and Schindelhauer, SPAA 2005]

Experimentally switch and flips protocol transform any graph in an expander very quickly.

#### Conjectures:

Switch converges on d-regular graph in O(nd) steps.

Flip converges on d-regular graph in  $O(nd \log n)$  steps.

# A new analysis for the two protocols

#### Results

Starting from any d-regular graph, with  $d \in \Omega(\log n)$ ,

the switch protocol transforms the graph in an algebraic expander in  $O\left(nd\right)$  steps.

the flip protocol transforms the graph in an algebraic expander in  $O\left(n^2d^2\sqrt{\log n}\right)$  steps.

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### **Obstacles**

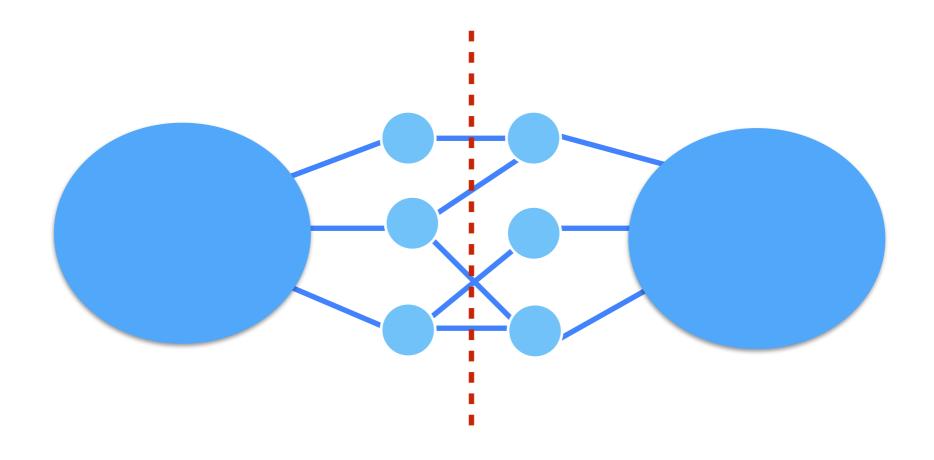
Dependencies.

Small cuts may first become smaller and only later increase.

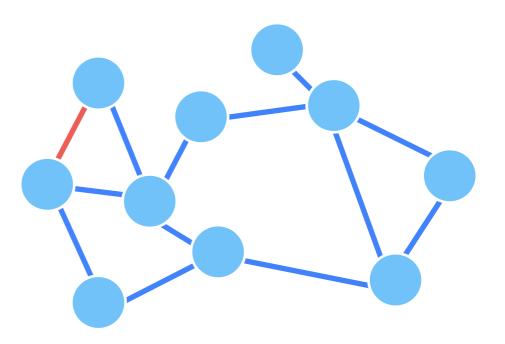
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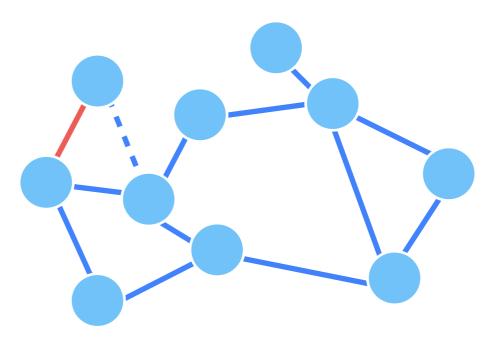


Pick a random edge.



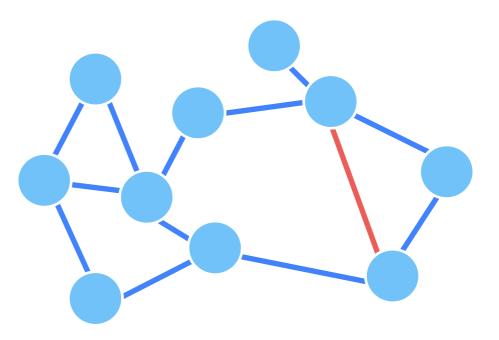
Pick a random edge.

One of the endpoints picks a neighbor at random(if in common, abort).



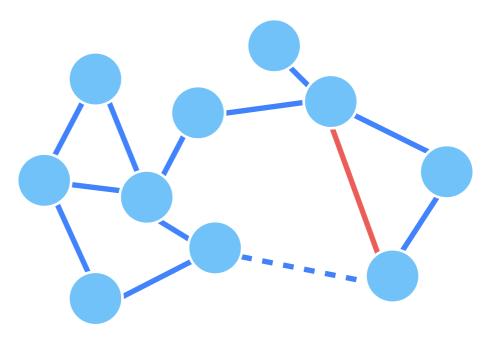
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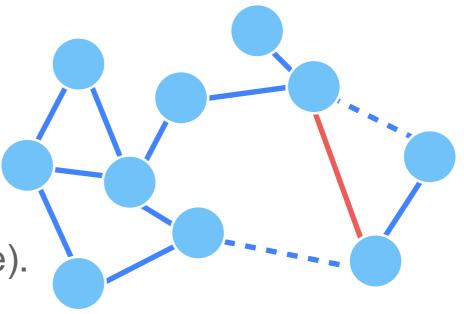
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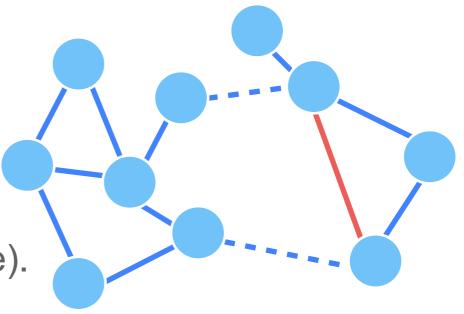
The other endpoint picks a random neighbor(if in common, picks a new one).



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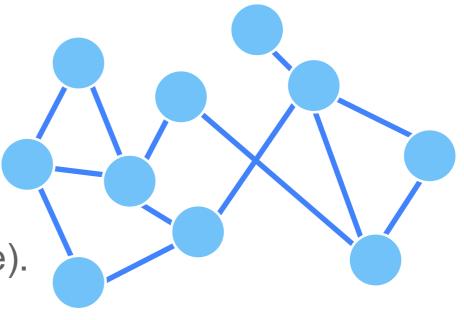
# Flip definition

Pick a random edge.

One of the endpoints picks a neighbor at random(if in common, abort).

The other endpoint picks a random neighbor(if in common, picks a new one).

Perform swap.

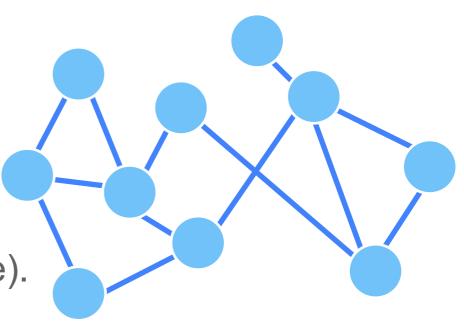


# Expected change of laplacian

Pick a random edge.

One of the endpoints picks a neighbor at random(if in common, abort).

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Perform swap.

Let 
$$\Delta^{(t)} = L\left(G^{(t+1)}\right) - L\left(G^{(t)}\right)$$

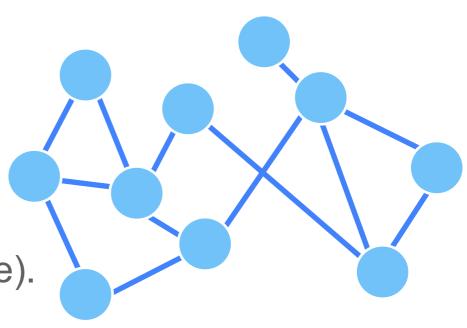
$$E\left[\Delta^{(t)}|G^{(t)}\right] = \frac{4}{d^2n} \left( (d+1)L^{(t)} - \left(L^{(t)}\right)^2 \right)$$

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Nice term.  $G^{(t)}$  has better expansion.

Unfortunately we cannot argue directly on the expectation of the matrix after t step.

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We use a classic potential used for matrix concentration:

$$\Phi^{(t)} = \hat{tr} \left( e^{-\frac{20\log n}{d} L^{(t)}} \right)$$

where 
$$\hat{tr}(e^A) = e^{\lambda_1} + e^{\lambda_2} + \dots$$

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Note that in order to have  $\Phi^{(t)}$  very small all the eigenvalues need to be large.

We want to show that the potential decreases

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$$by e^{-A} = I - A + A^2$$

$$= \hat{tr} \left( e^{-\frac{20\log n}{d}L^{(t)}} \left( I - \frac{20\log n}{d} \Delta^{(t)} + \left( \frac{20\log n}{d} \Delta^{(t)} \right)^2 \right) \right)$$

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$$E\left[\Phi^{(t+1)}|G^{t}\right] = \Phi^{(t)} - \frac{4\log n}{d^{3}n}\hat{tr}\left(e^{-\frac{20\log n}{d}L^{(t)}}\left(L^{(t)}\left(\frac{d}{2}\hat{I} - L^{(t)}\right)\right)\right)$$

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Using common diagonalization

$$\sum_{1 \le i \le n} e^{-\frac{20 \log n}{d} \lambda_i} \lambda_i (d/2 - \lambda_i)$$

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Two interesting cases:

$$\forall i: \lambda_i \ge \frac{d}{4}$$

$$\sum_{1 \le i \le n} e^{-\frac{20 \log n}{d} \lambda_i} \lambda_i (d/2 - \lambda_i) \in O(n^{-3})$$

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Two interesting cases:

$$\exists i: \lambda_i < \frac{d}{4}$$

We look at:

$$\frac{\sum_{1 \le i \le n} e^{-\frac{20 \log n}{d} \lambda_i} \lambda_i (d/2 - \lambda_i)}{\Phi^{(t)}} = \frac{\sum_{1 \le i \le n} e^{-\frac{20 \log n}{d} \lambda_i} \lambda_i (d/2 - \lambda_i)}{\sum_{1 \le i \le n} e^{-\frac{20 \log n}{d} \lambda_i}}$$

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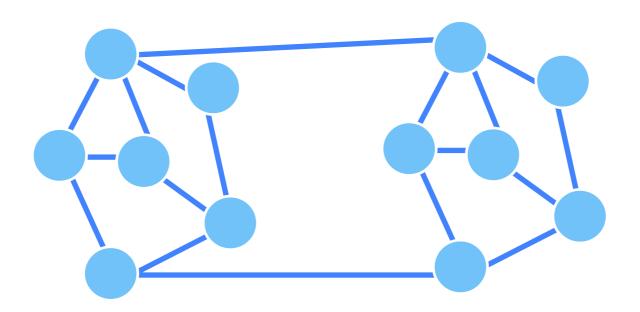
Thus:

$$E\left[\Phi^{(t+1)}|G^{(t)}\right] = \left(1 - \Omega\left(\frac{\sqrt{\log n}}{n^2 d^2}\right)\right)\Phi^{(t)} + O(n^{-3})$$

So in expectation  $\Phi^{(t)}$  is in  $O(n^{-3})$  after  $O(n^2d^2\log n)$  steps, hence using Markov inequality we get the result.

# Limit of our analysis

Expected additive improvement in a round can be  $O\left(\frac{1}{n^2d^2}\right)$ 



# Conclusions and future directions

### Conclusions

New technique to analyze distribute protocol

New convergence time analysis for flip and switch protocol

## **Future works**

Improve analysis of the flip

Study parallelized version of the protocol

Study node addition or deletion

# Thanks!