Causal Limits of Distributed Computation



Francesco d'Amore

Based on the works [Coiteux-Roy et al. STOC '24], [Akbari et al. STOC '25], [Balliu et al. STOC '25], [Balliu et al. '25].

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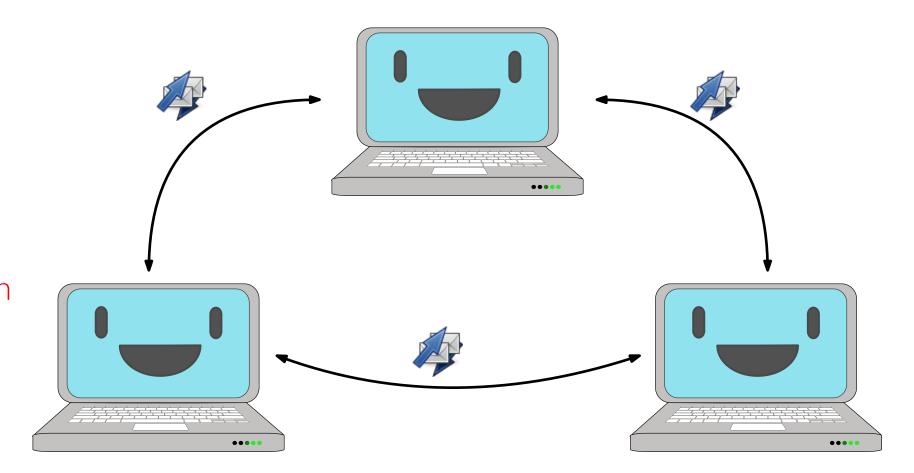
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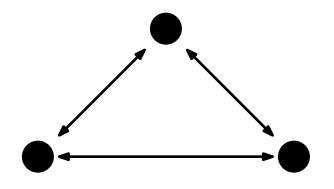
1. Intro

- Distributed computation
- The LOCAL model of computation
- Locally checkable labeling (LCL) problems
- 2. Quantum and causality-based models
- The non-signaling model & bounded-dependence model
- State-of-the-art lower bounds & upper bounds
- 3. From quantum back to classical
- Simulation in weaker models
- 4. Conclusions and open problems

• Synchronous distributed network

- graph G = (V, E) with |V| = n
- E: communication links
- discrete communication rounds: t = 1, 2, ...
- each node in V runs the same algorithm
- each round = local computation + communication



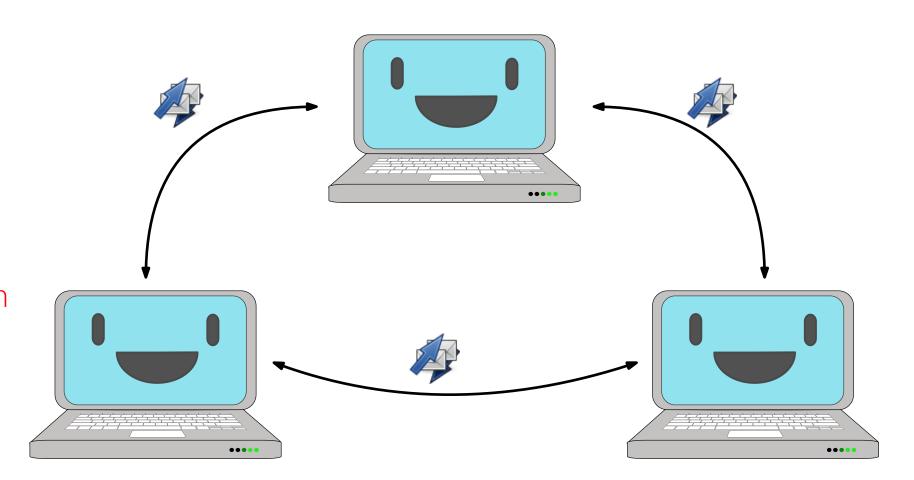


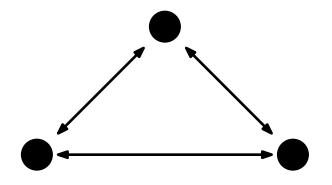
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Problem examples

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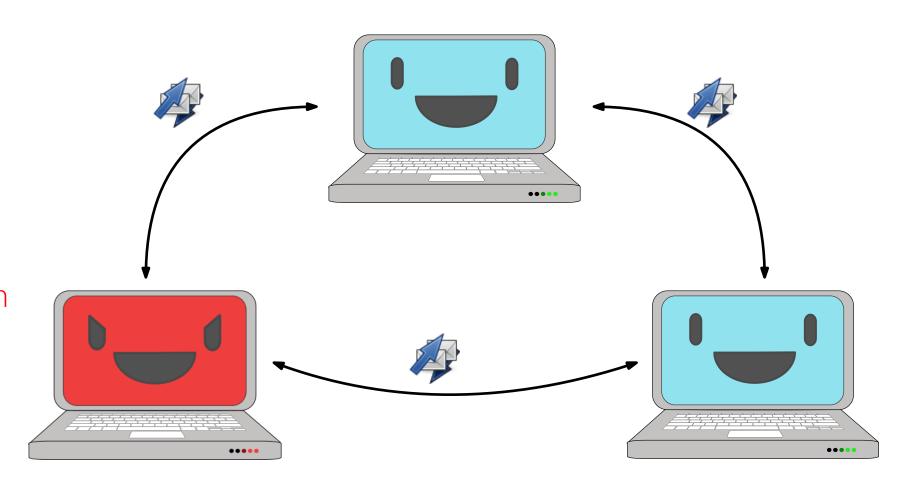


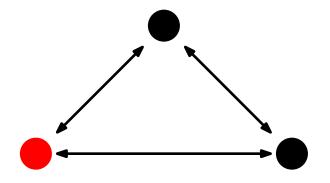
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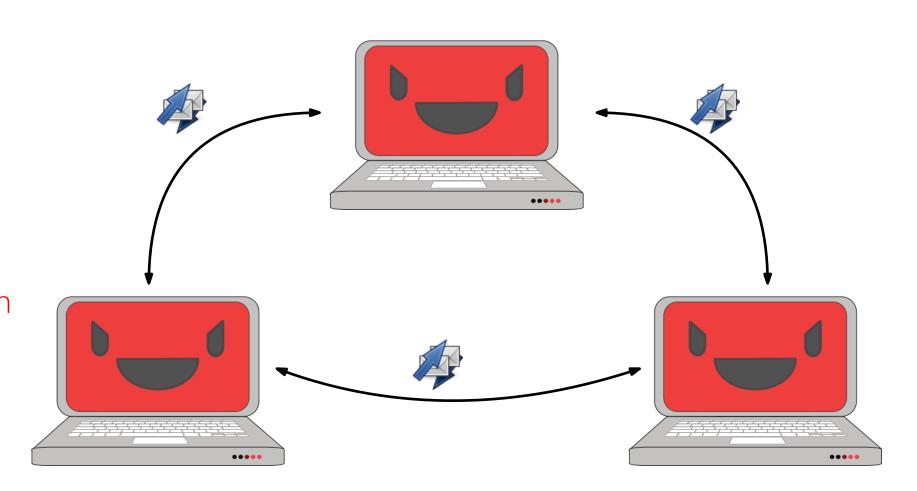


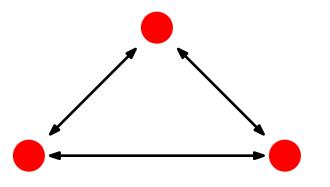
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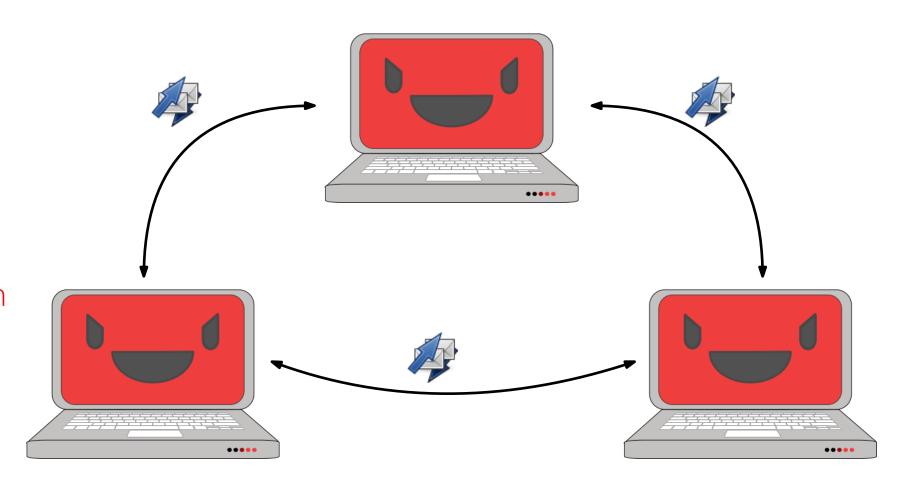


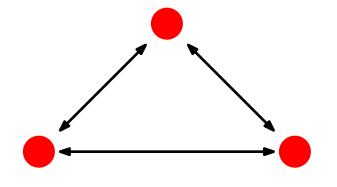
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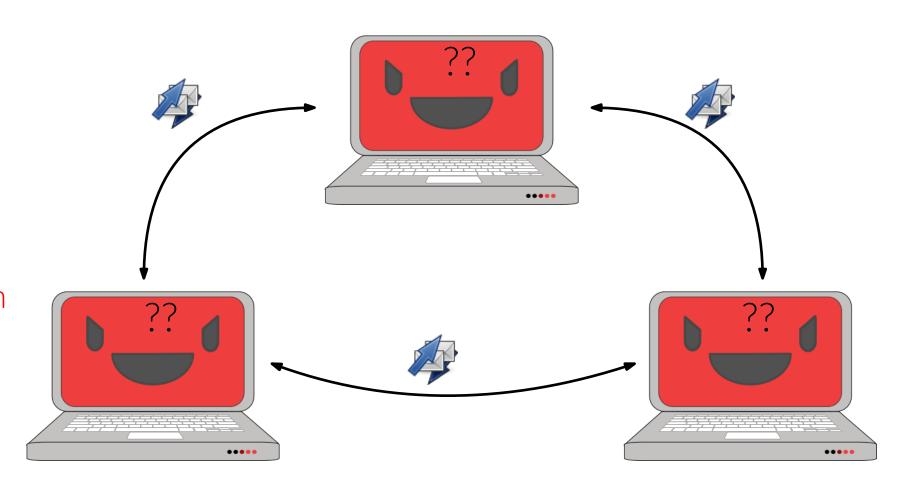


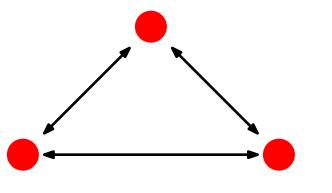
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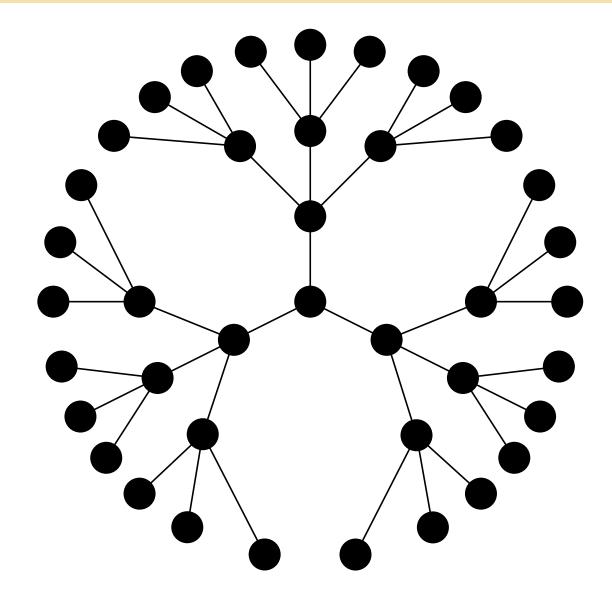
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 - algorithm: ?? we need to break symmetry

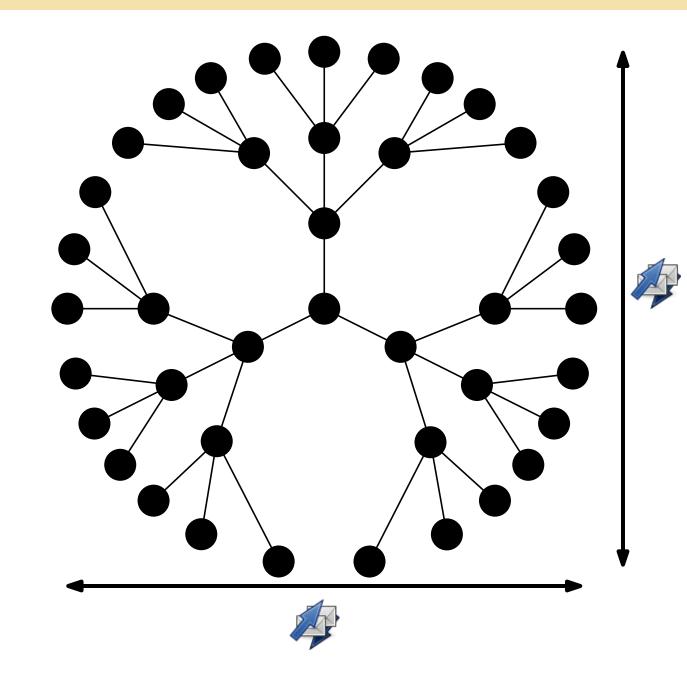




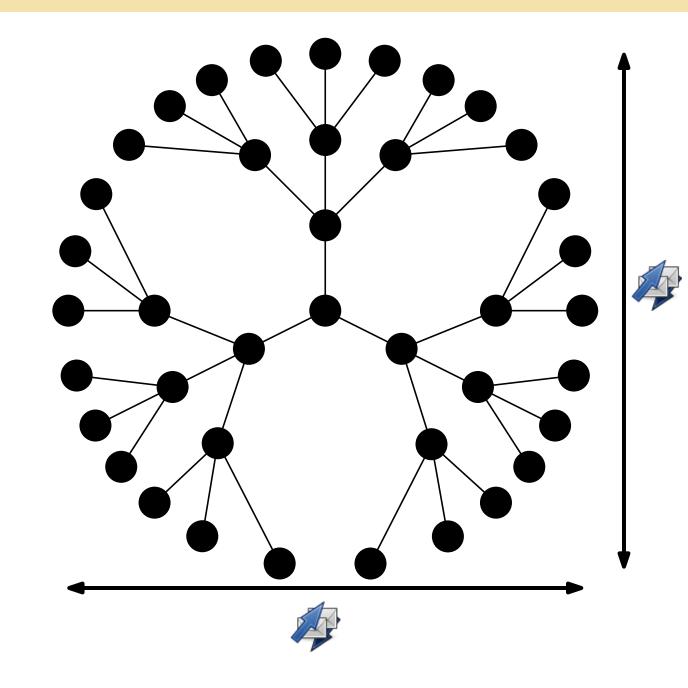
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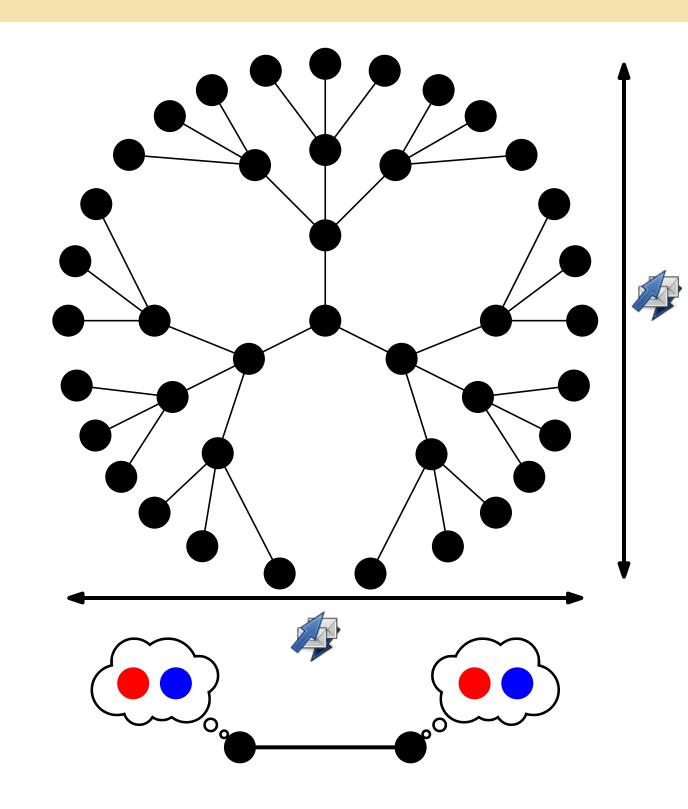
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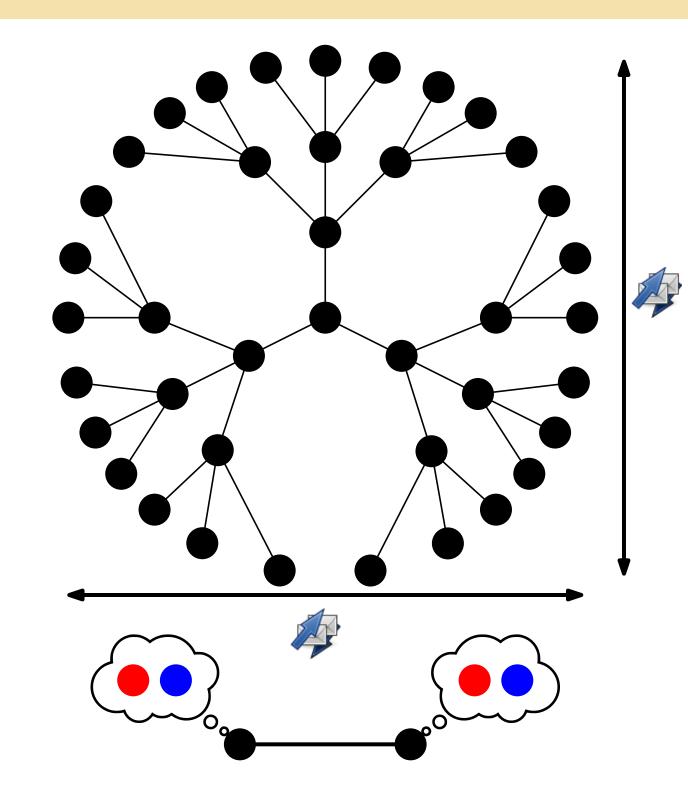
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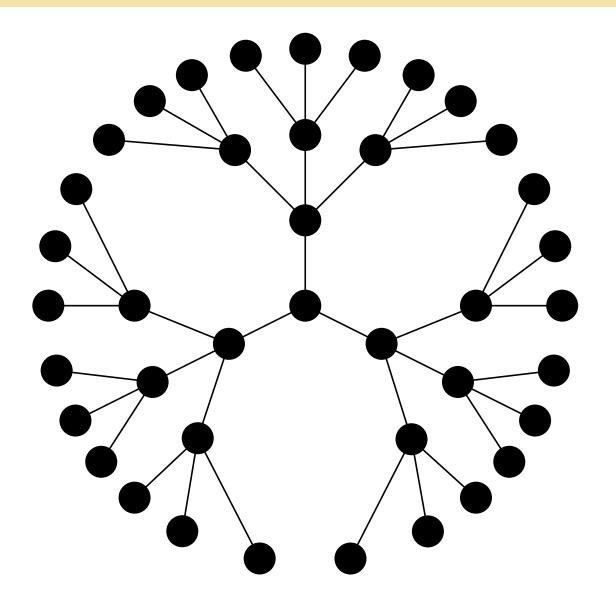
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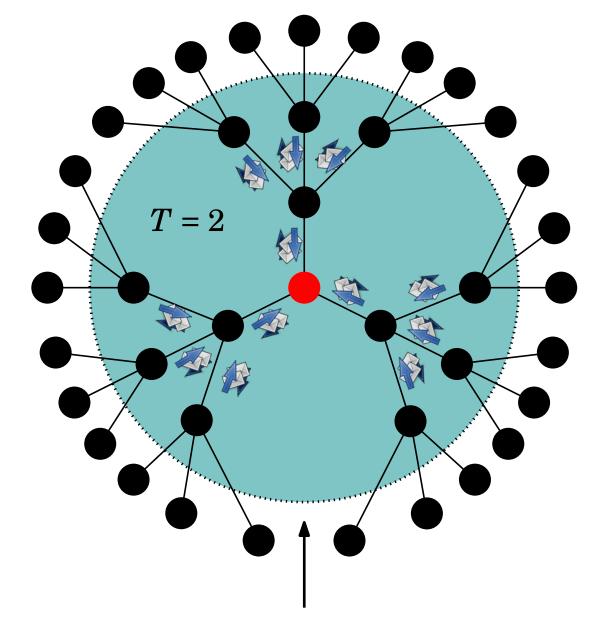
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Complexity measure: number of communication rounds

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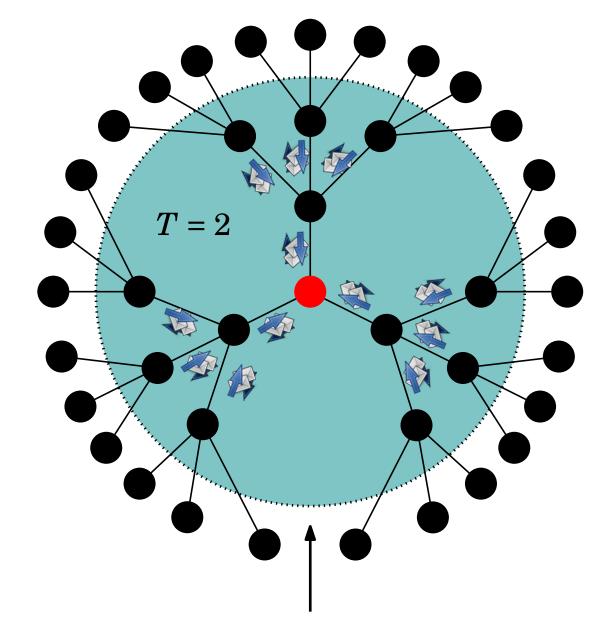


knowledge after 2 rounds of communication

Complexity measure: number of communication rounds

Equivalence:

- A: T(n)-round LOCAL algorithm
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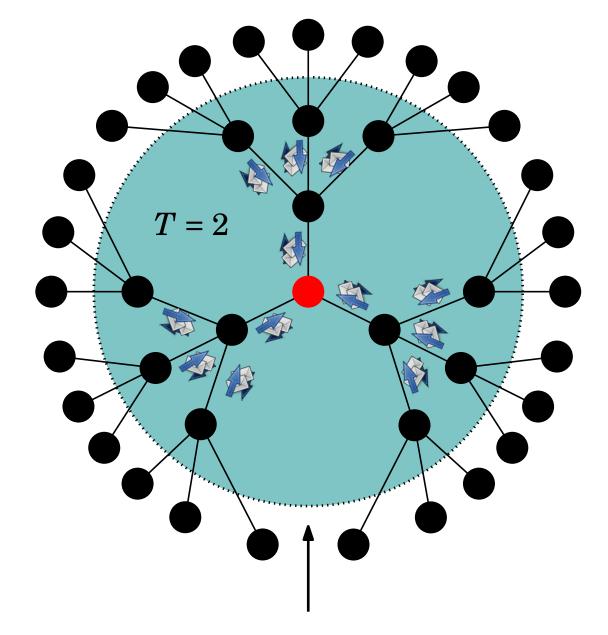
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- given a LOCAL algorithm A, we can construct the mapping f
- given f, we can construct a LOCAL algorithm B that simulates f



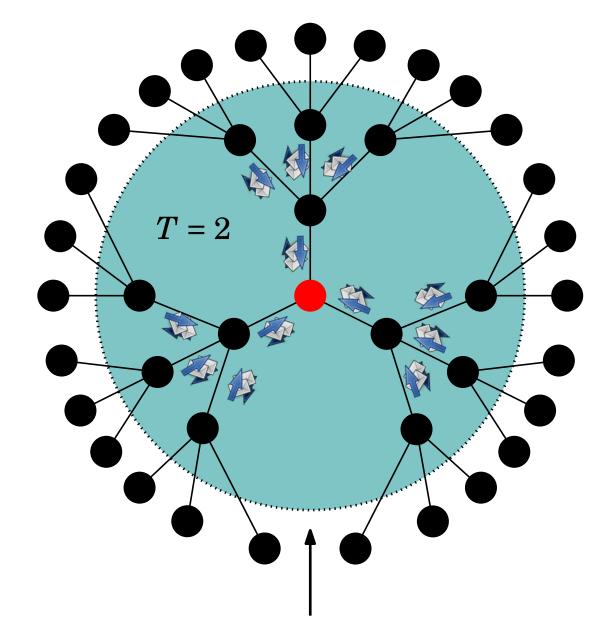
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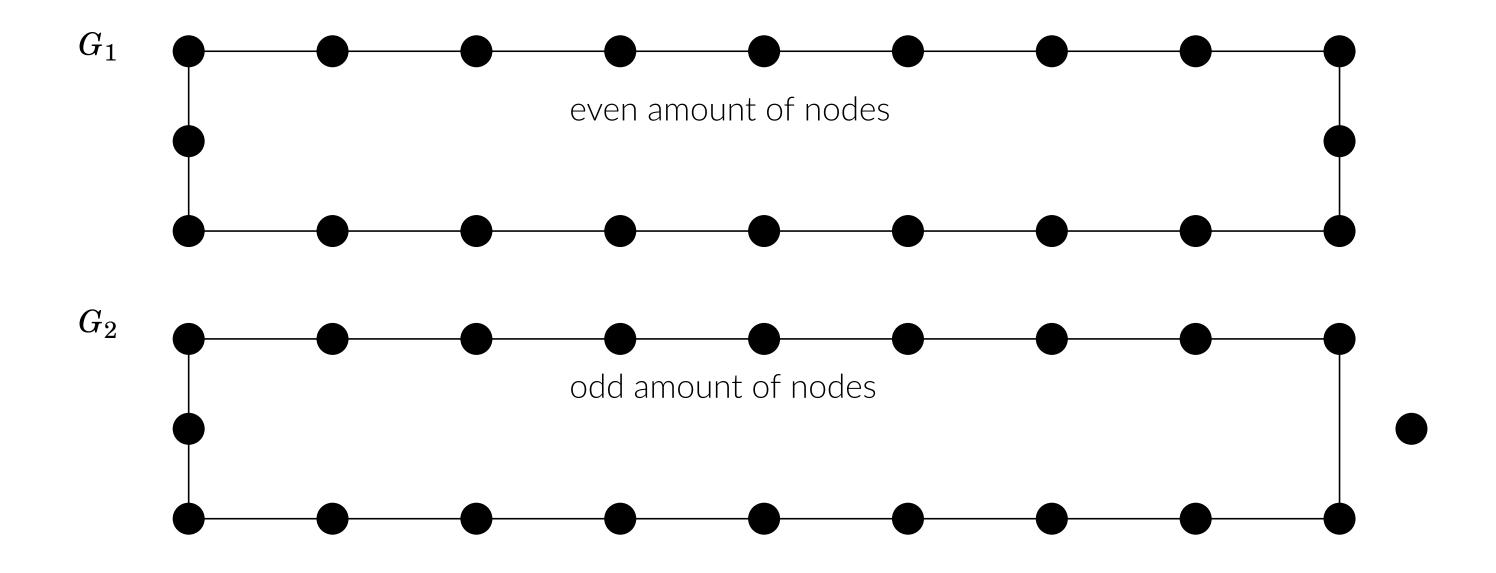
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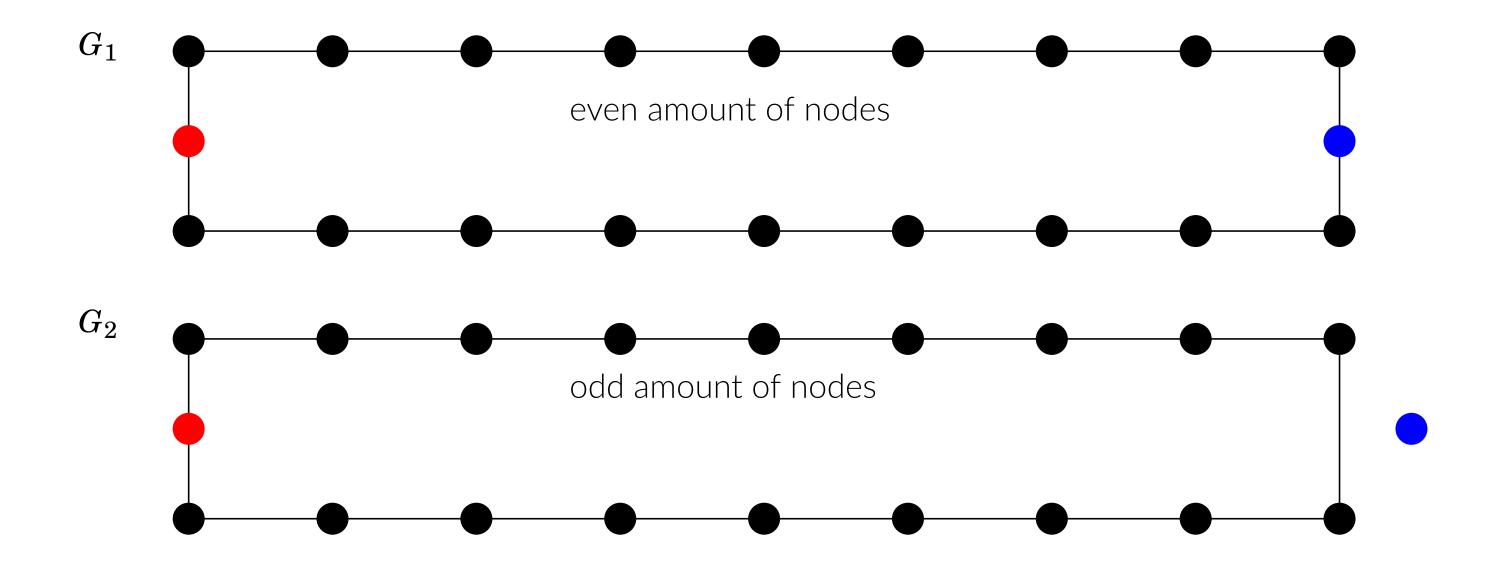
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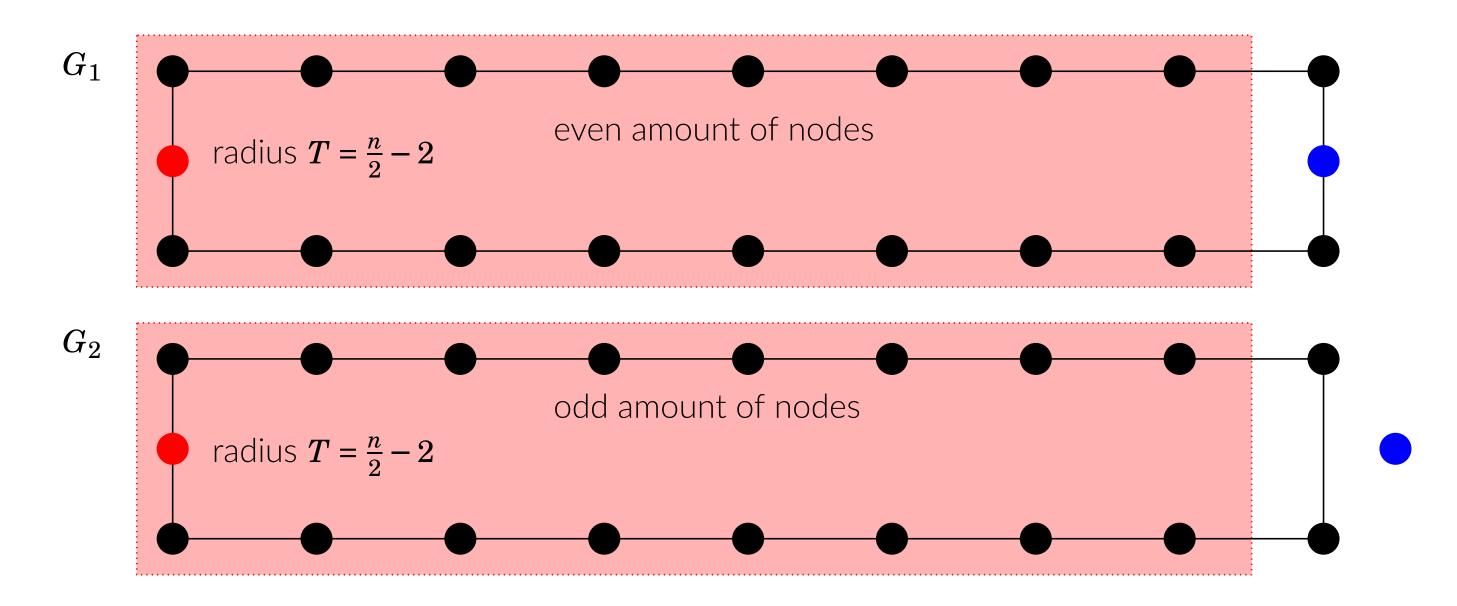


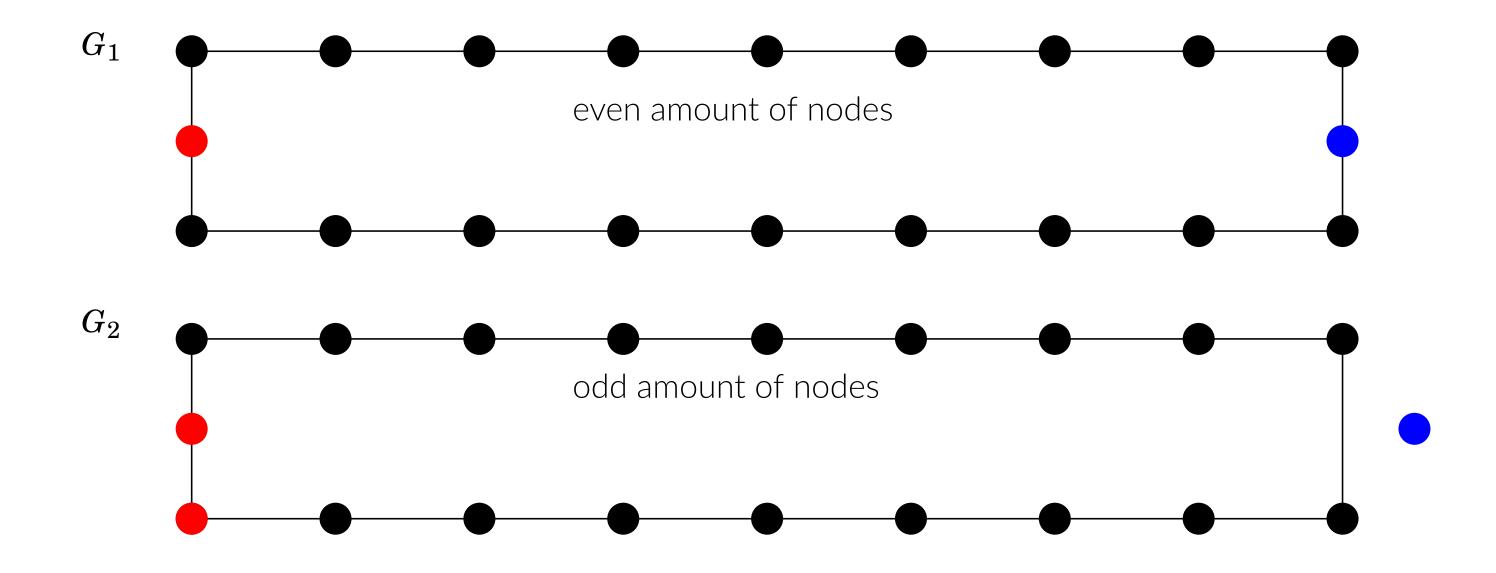
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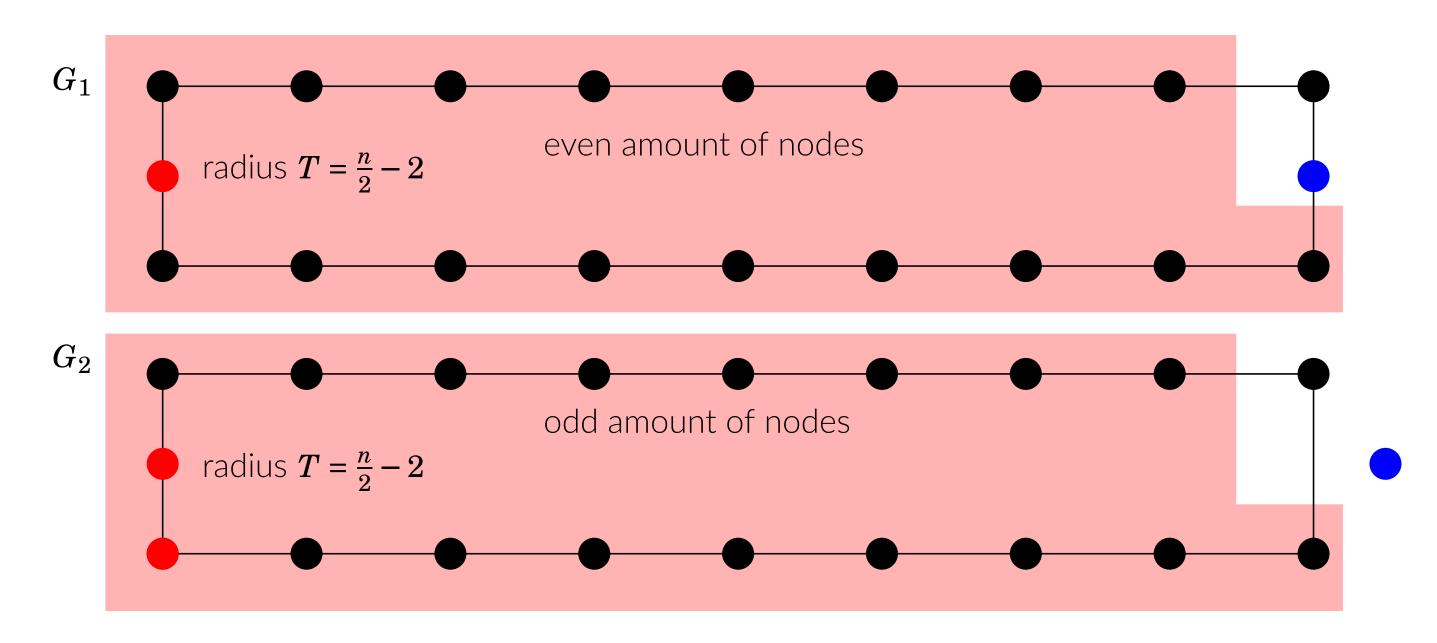
• Locality T = diam(G) + 1 is always sufficient to solve any problem: gathering algorithm

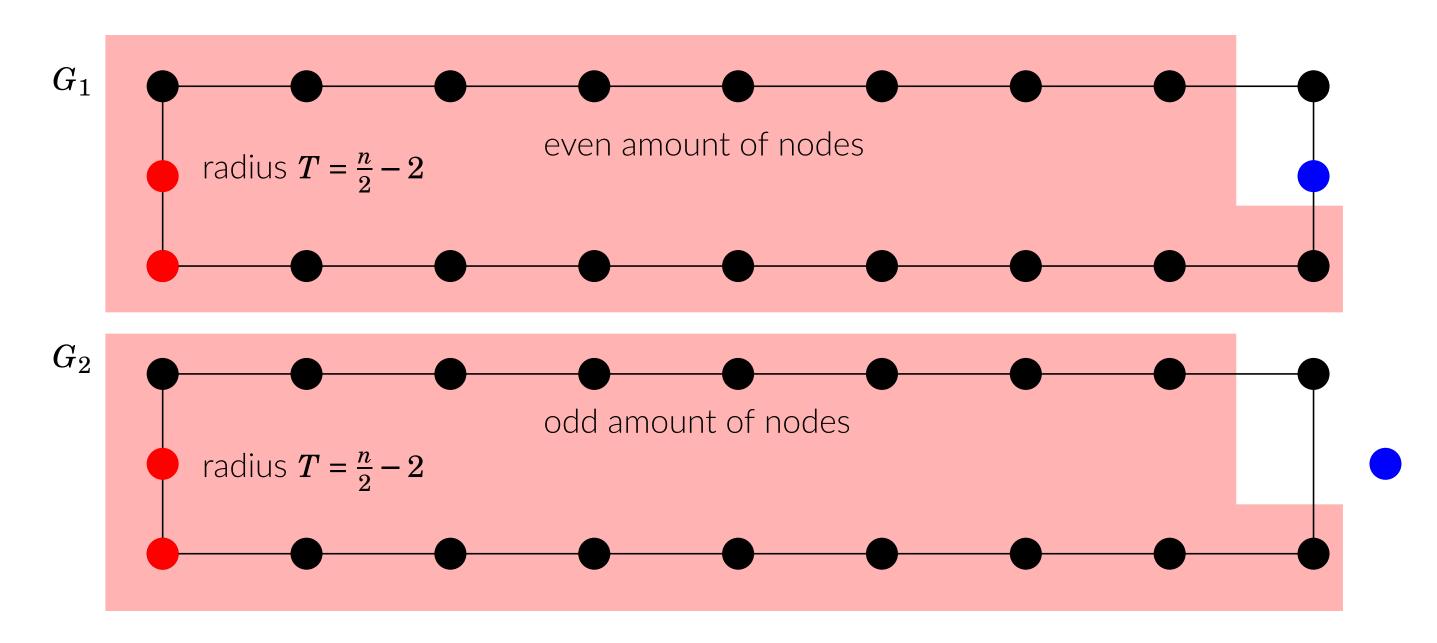




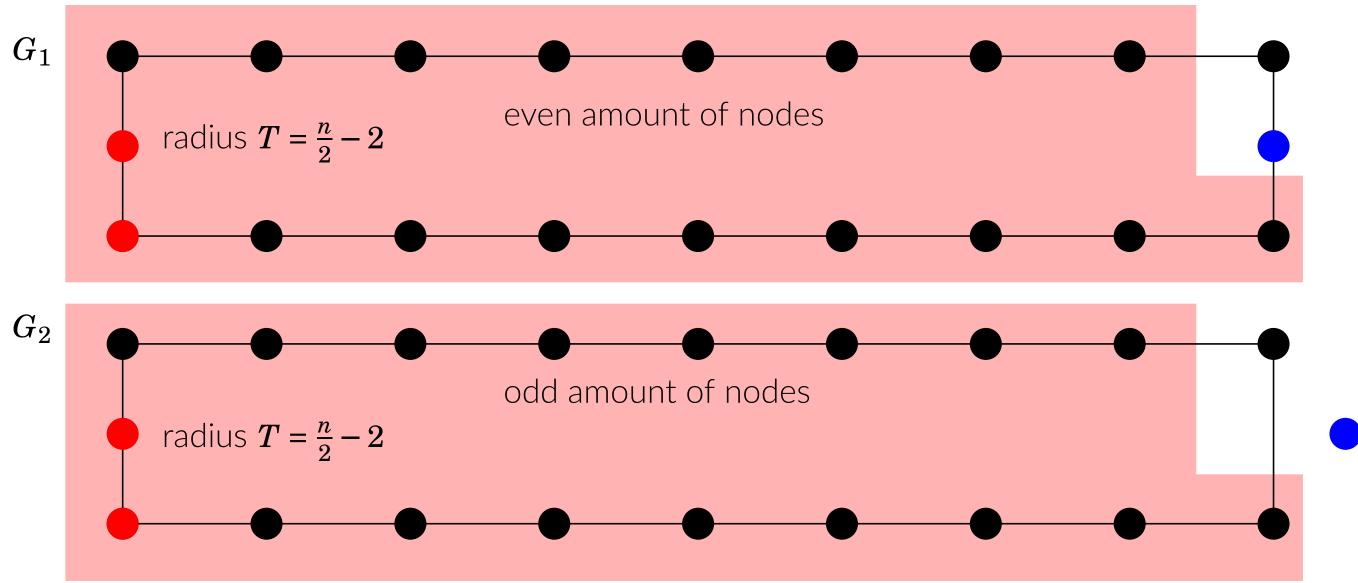








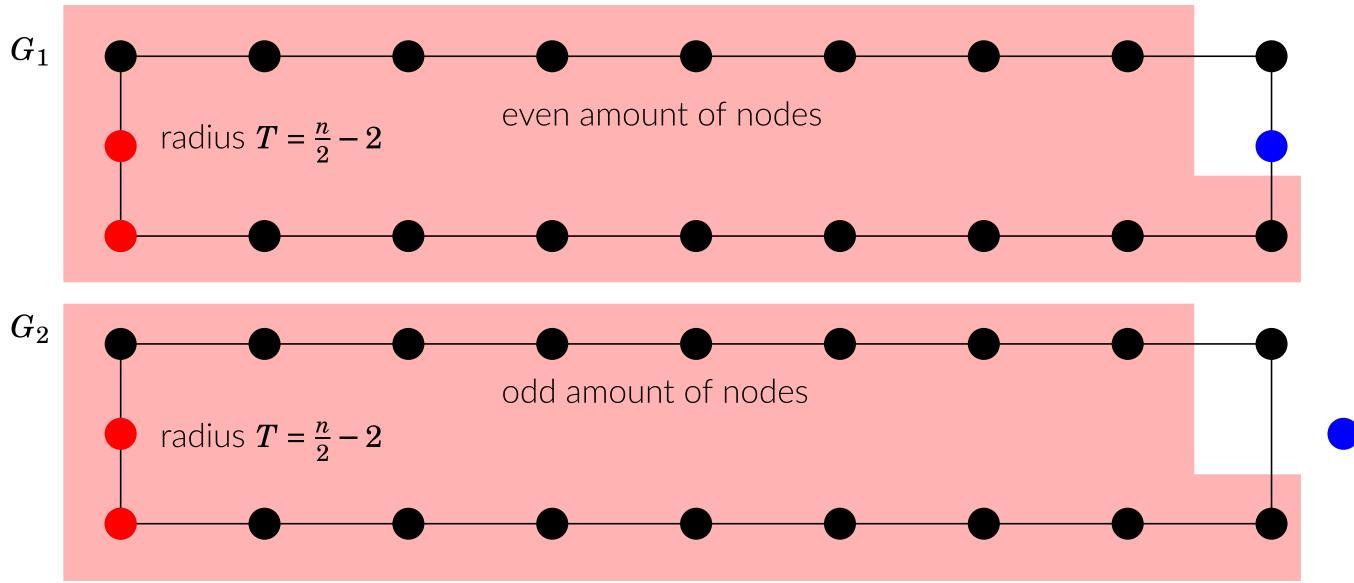
• **Problem**: 2-coloring even cycles. Assume we have a T-round LOCAL algorithm with $T \leq \frac{n}{2} - 2$





existence of o(n)-round LOCAL algorithm that 2-colors even cycles \implies 2-coloring of odd cycles

• **Problem**: 2-coloring even cycles requires locality $\Omega(n)$



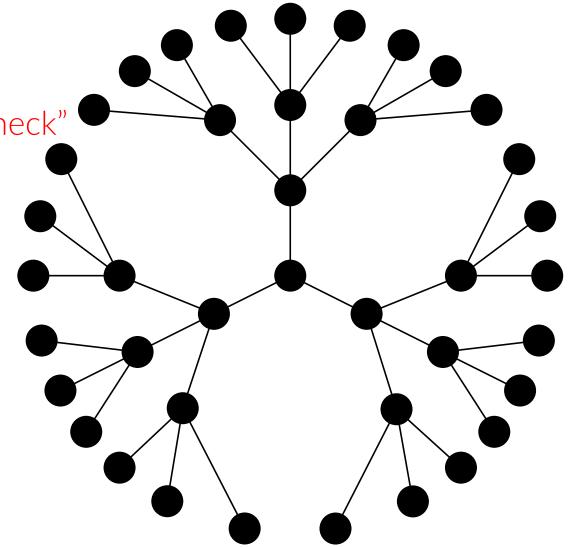


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[Naor and Stockmeyer STOC '93 & SICOMP '95]

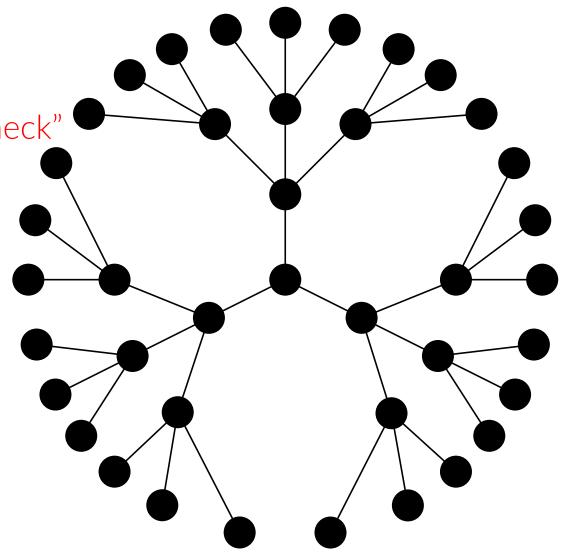
• Problems whose solutions might be "hard to find" but are "easy to check"

- "analogue" of NP in the distributed setting
- coloring, maximal independent set, maximal matching, etc.



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 - radius $r = \Theta(1)$
 - each node can check its solution within its radius-r neighborhood
 - a globally valid iff each node is locally happy

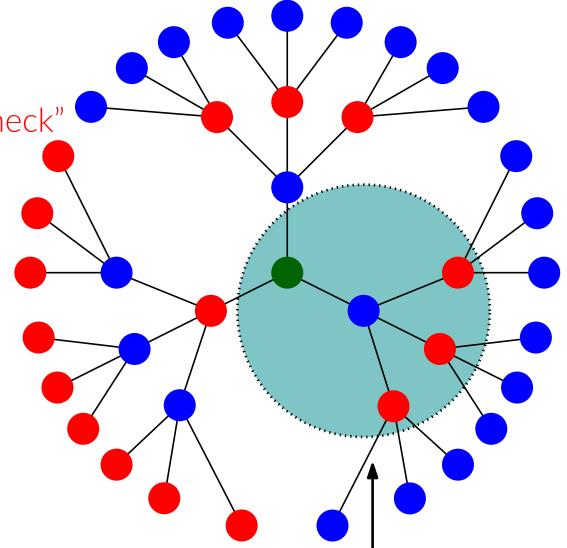


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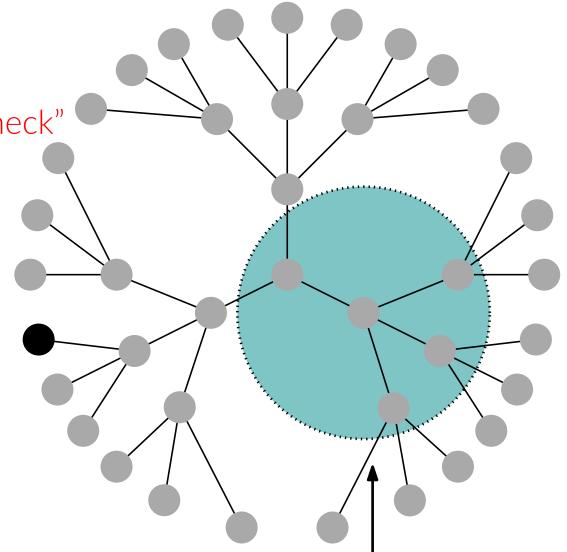
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Leader election: the checking radius should be r = diam(G)

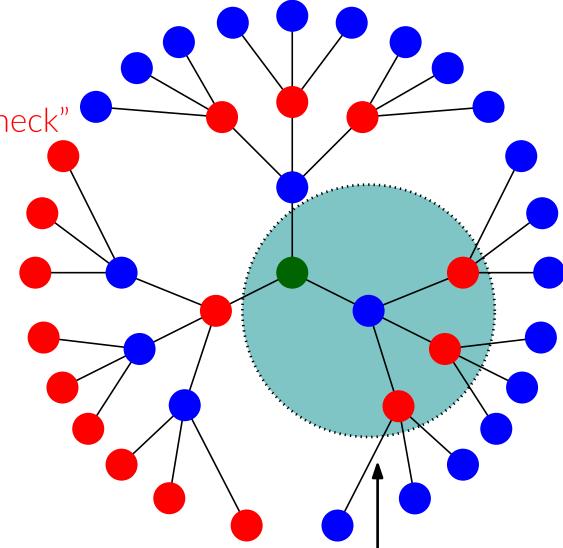
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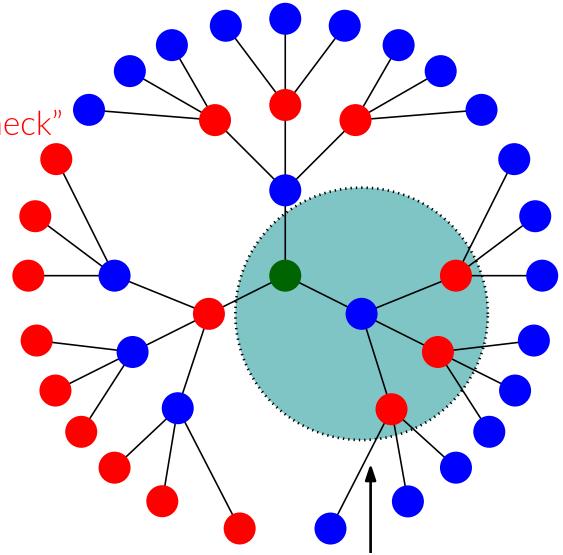
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 - classification of LCLs based on complexity (locality)
 - e.g.: complexity T(n) in randomized-LOCAL $\Longrightarrow O(T(2^{n^2}))$ in deterministic-LOCAL [Chang et al. SICOMP '19]



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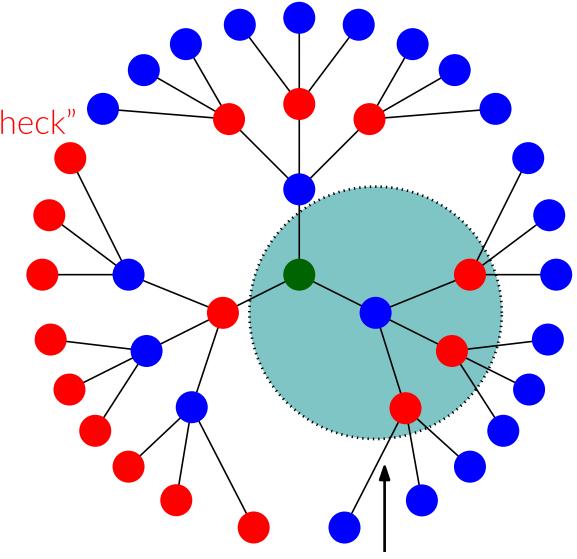
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 - [BFHKLRSU STOC '16; BHKLOPRSU PODC'17; GKM STOC '17; GHK FOCS '18; CP SICOMP '19; BHKLOS STOC '18; BBCORS PODC '19; BBOS PODC '20; BBHORS JACM '21; BBCOSS DISC '22; AELMSS ICALP '23; etc.]



3-coloring: the blue node checks if its color is different from those of its neighbors

Complexity landscape of LCLs

Paths and cycles

det-LOCAL	O(1)	$\Theta(\log^\star n)$	$\Theta(n)$	
rand-LOCAL	rand-LOCAL $O(1)$		$\Theta(n)$	

ullet Balanced d-dimensional toroidal grids

det-LOCAL	O(1)	$\Theta(\log^\star n)$	$\Theta(n^{1/d})$	
rand-LOCAL	d-LOCAL $O(1)$		$\Theta(n^{1/d})$	

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• Bounded-degree trees

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General graphs??

• Bounded-degree trees

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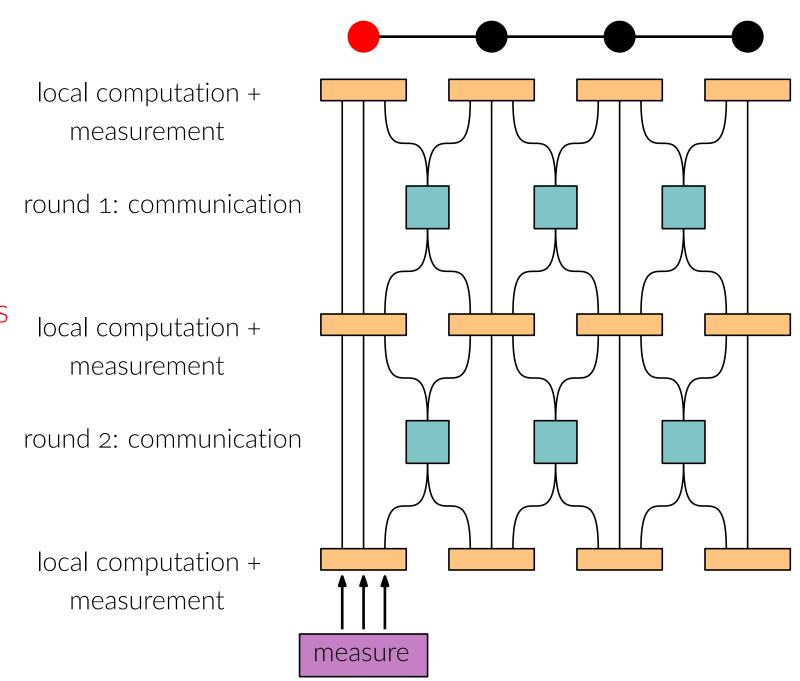
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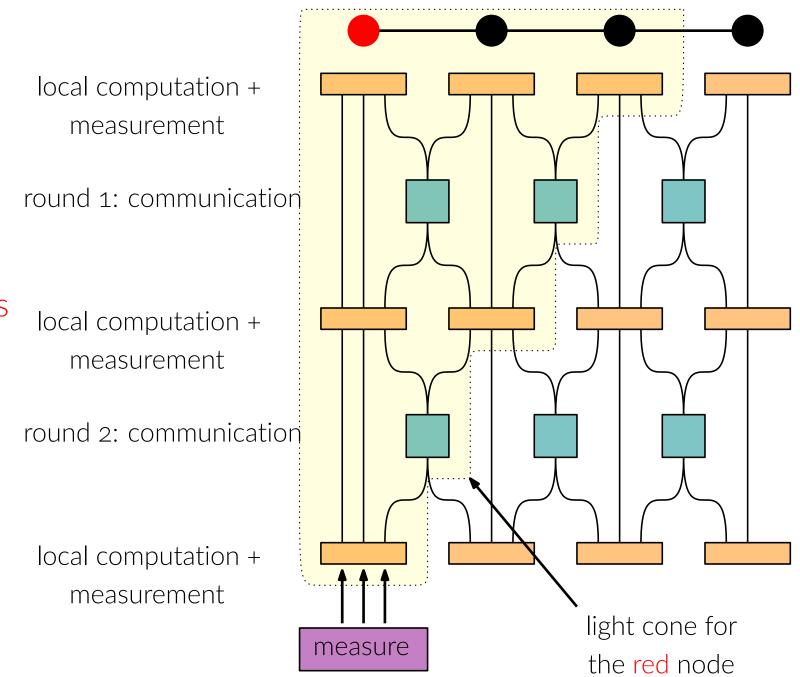
[Gavoille et al. DISC '09]

- Distributed system of n quantum processors/nodes
 - quantum computation
 - quantum communication (qubits)
 - output: measurement of qubits
- Complexity measure: number of communication rounds



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local computation + measurement round 1: communication local computation + measurement round 2: communication local computation + measurement measure

• **Question**: is there any graph problem that admits quantum advantage?

light cone for

the red node

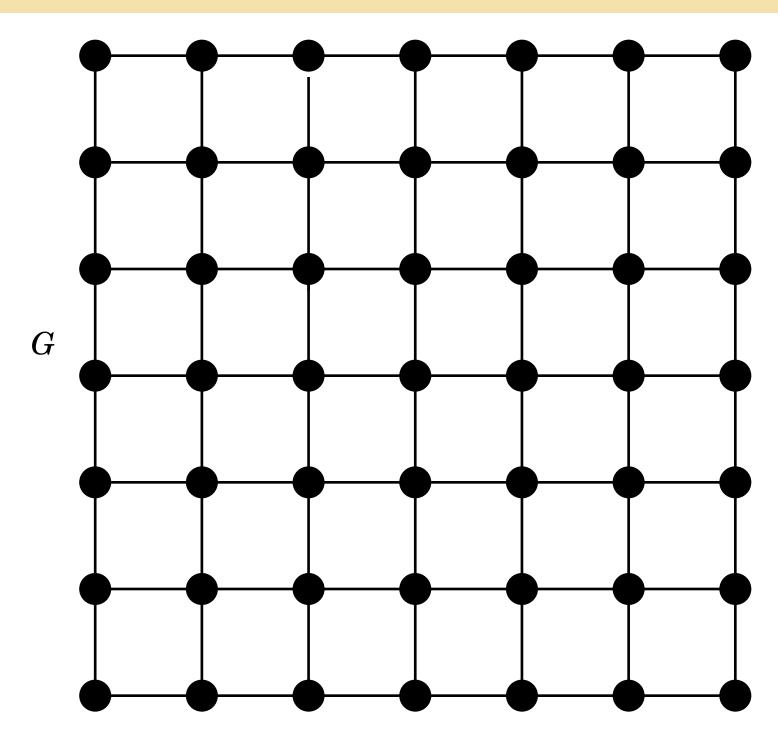
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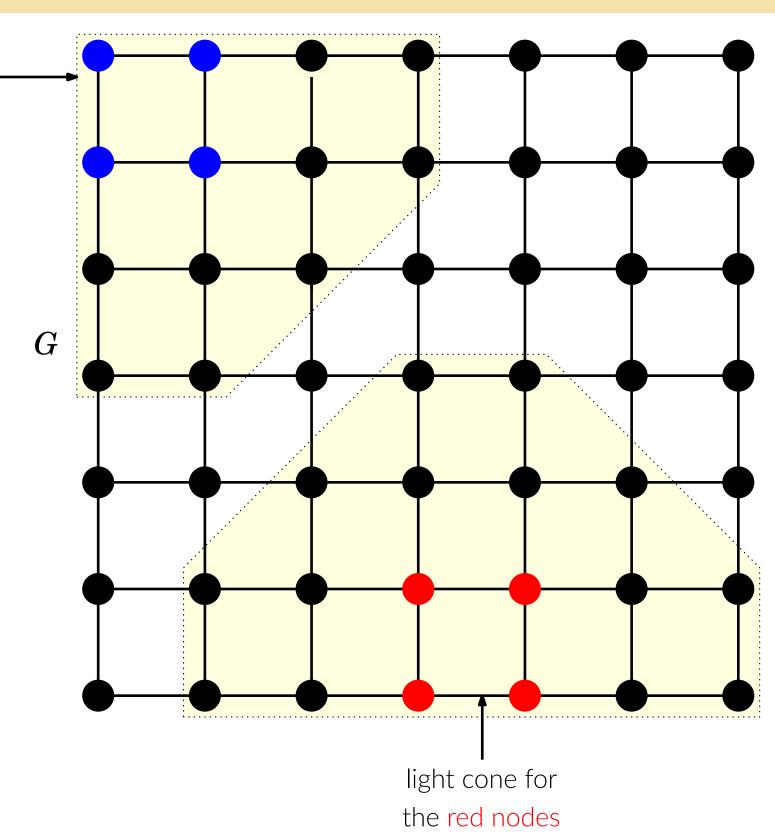
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 - Yes! But artificial LCLs [Balliu et al. STOC '25; Balliu et al. '25]
 - Otherwise we do not know
- What do we know?
 - focus on LCLs
 - input graph degree is bounded by a constant Δ [Naor and Stockmeyer SICOMP '95]

ullet Run a 2-round algorithm A in G



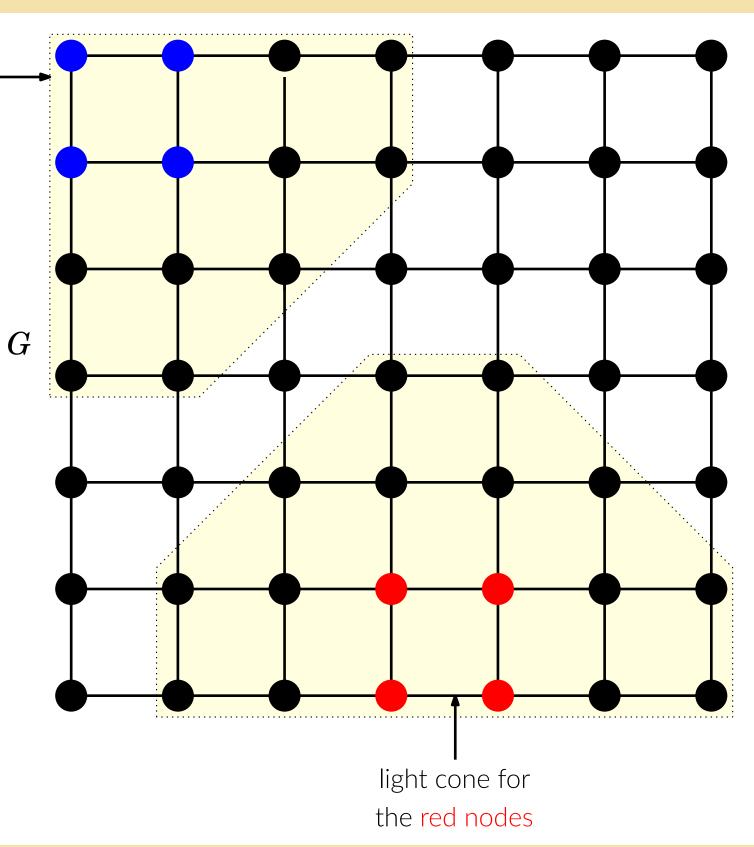
light cone for

- Run a 2-round algorithm A in G
 - output for the red and blue nodes is determined by their respective light cones



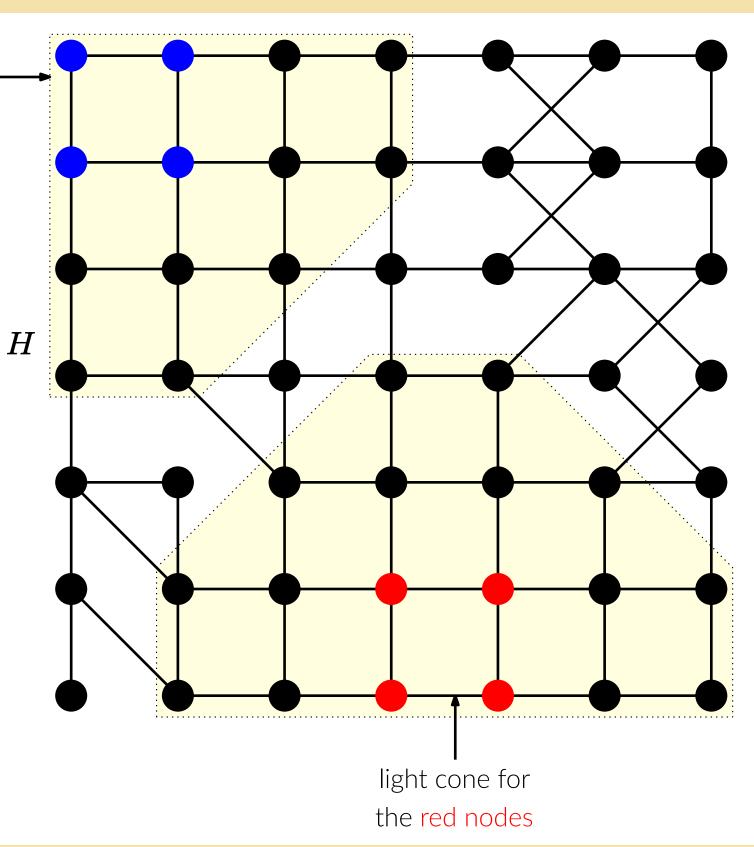
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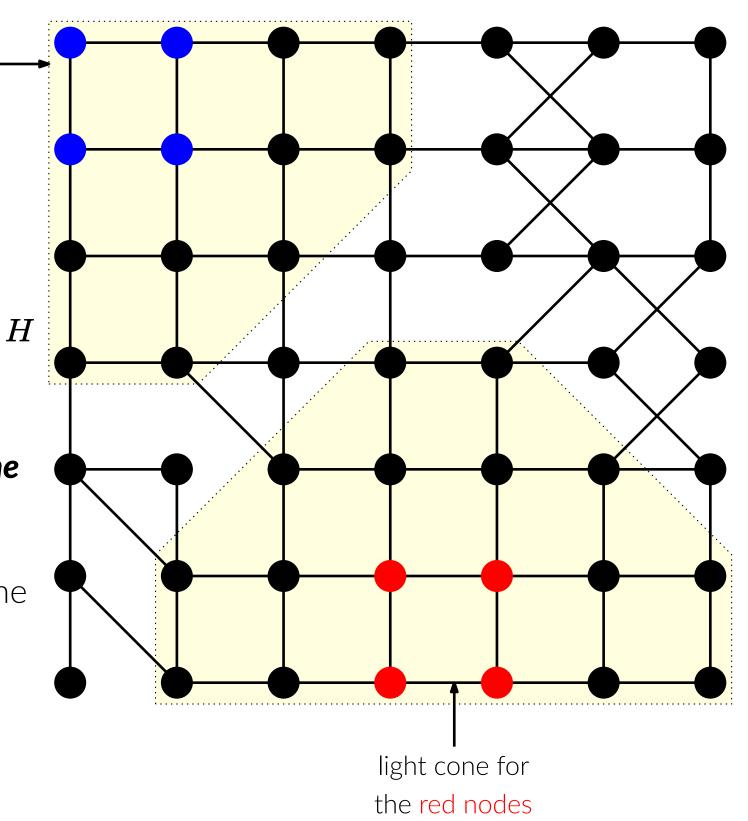
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- Output distributions for red and blue nodes are independent
 - as long as their distance is at least 5
- Output distributions remains the same if light cone is the same
 - non-signaling property
 - changes that are beyond 2-hops away do not influence the output distribution
 - also known as causality



Abstracting output distributions

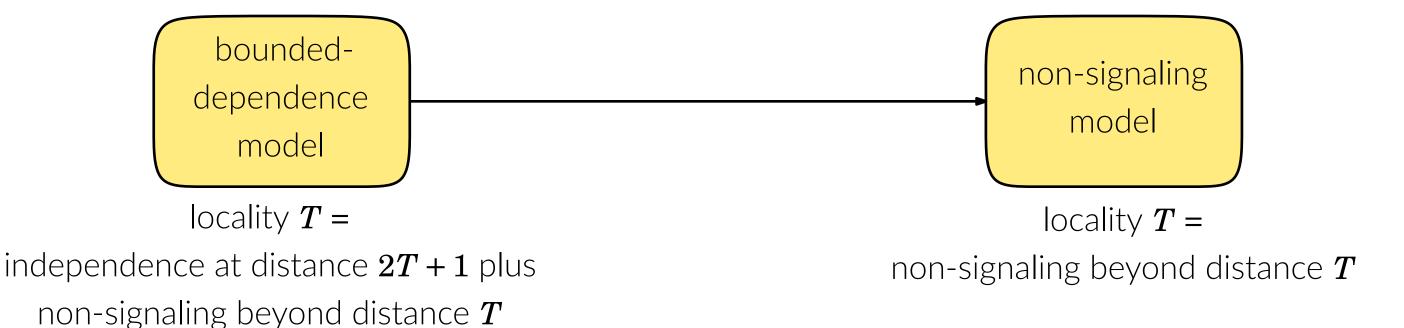
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 - outputs of subsets of nodes at distance more than $\mathbf{2}T$ are independent
 - non-signaling beyond distance $oldsymbol{T}$
- Then we can just think about output distributions!
 - computational models that produce directly distributions with the aforementioned properties

Abstracting output distributions

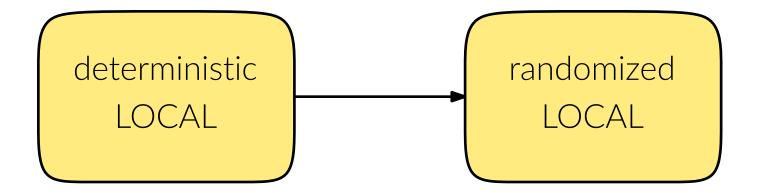
- \bullet A T-round distributed algorithm yields an **output distribution** with the following **properties**:
 - outputs of subsets of nodes at distance more than 2T are independent
 - non-signaling beyond distance $oldsymbol{T}$
- Then we can just think about output distributions!
 - computational models that produce directly distributions with the aforementioned properties



[Holroyd and Liggett Forum of Mathematics Pi '14] [Akbari et al. STOC '24]* finitely-dependent distributions if T = O(1)

[Gavoille et al. DISC '09] [Arfaoui and Fraigniaud PODC '12 & SIGACT News '14]

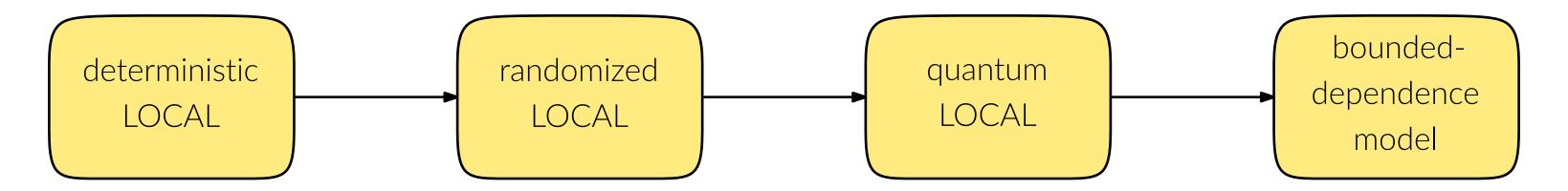
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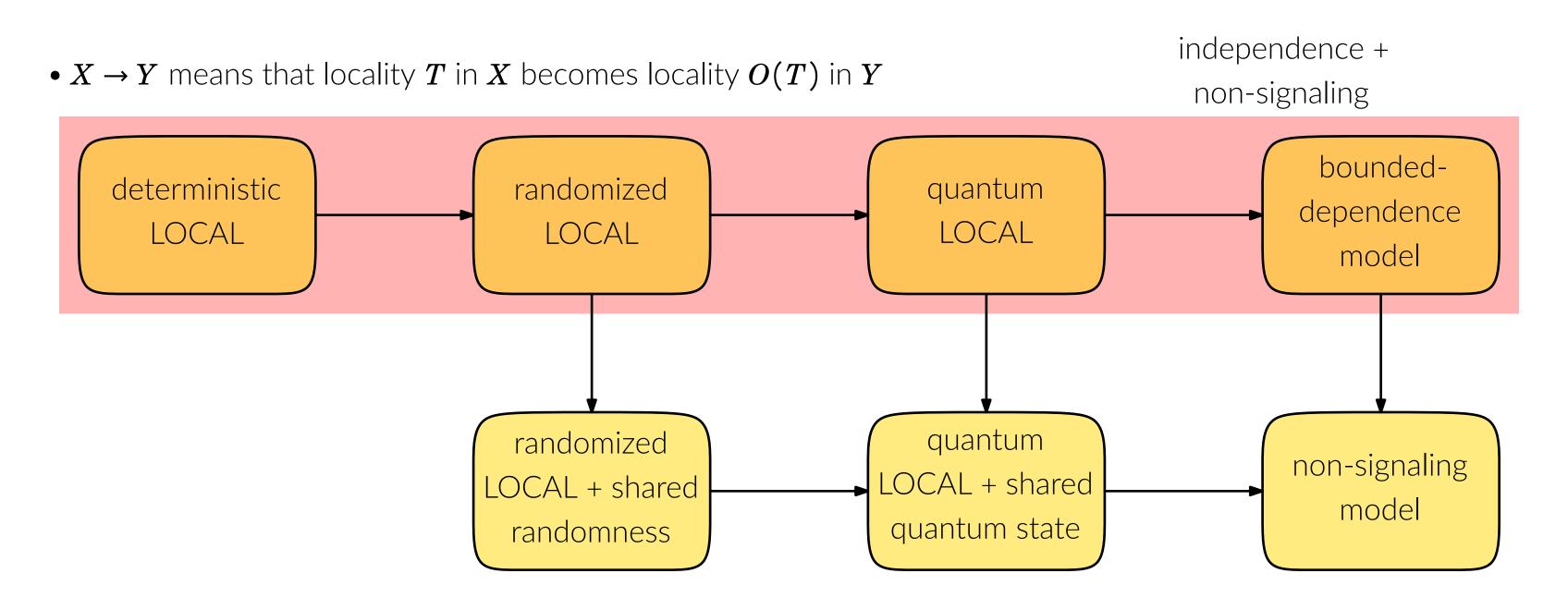
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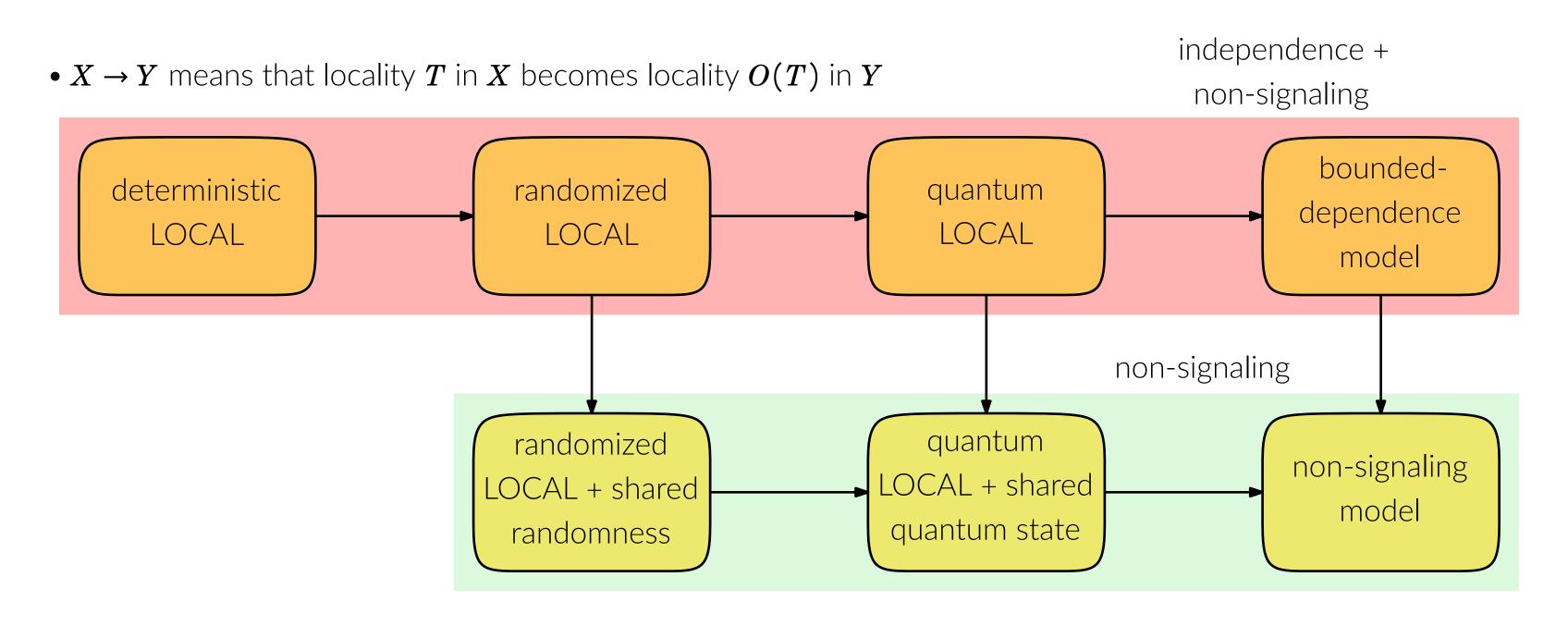


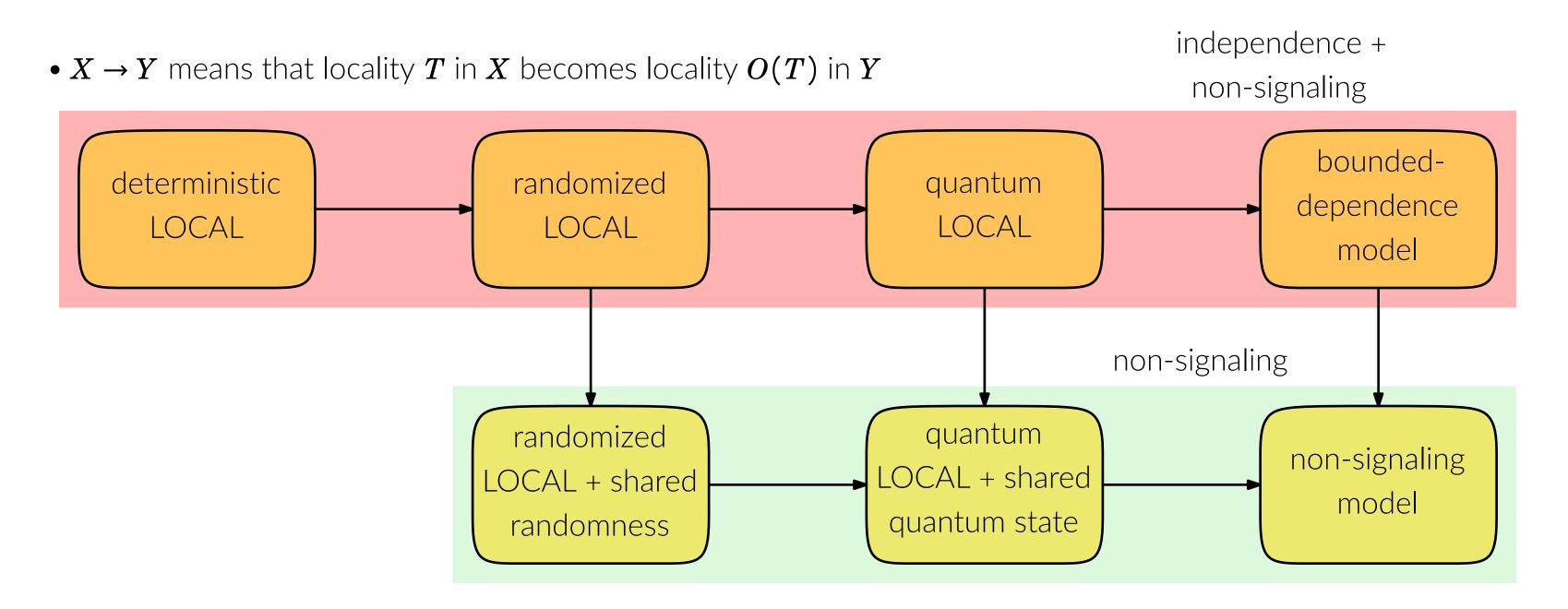
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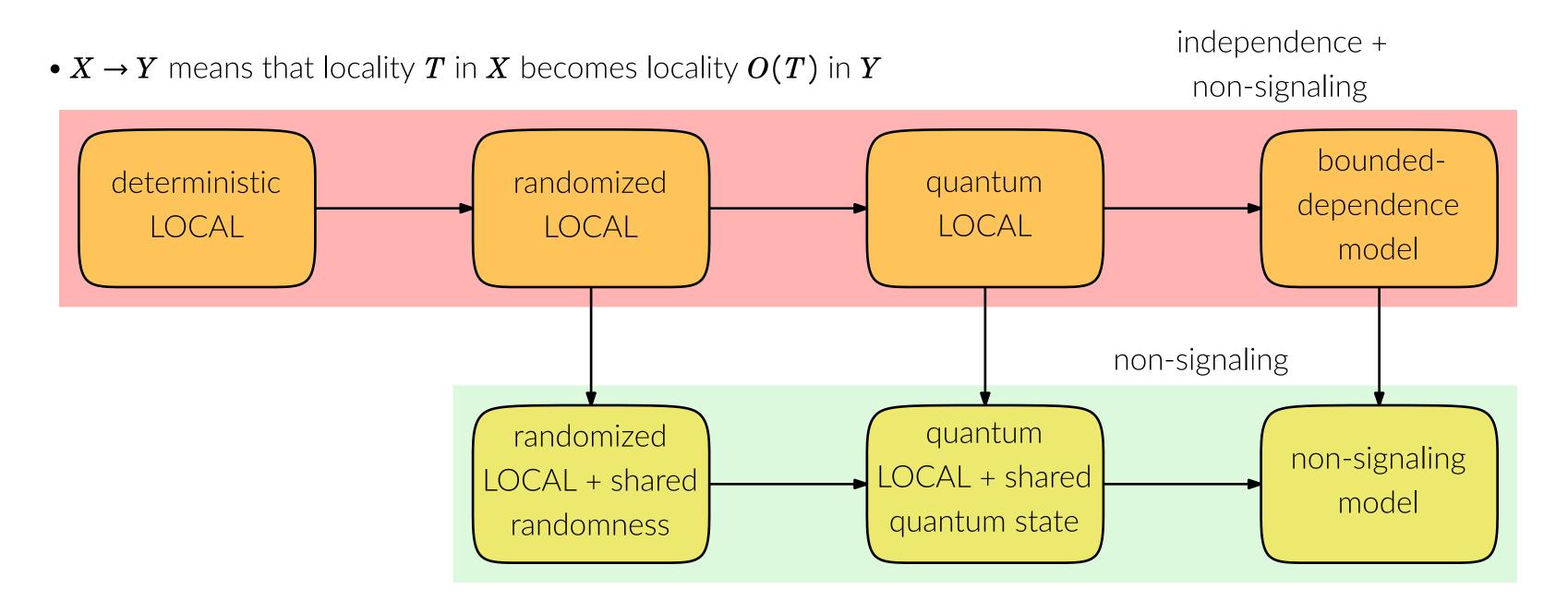
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• Is it possible to "sandwich" quantum-LOCAL between weaker and stronger models?

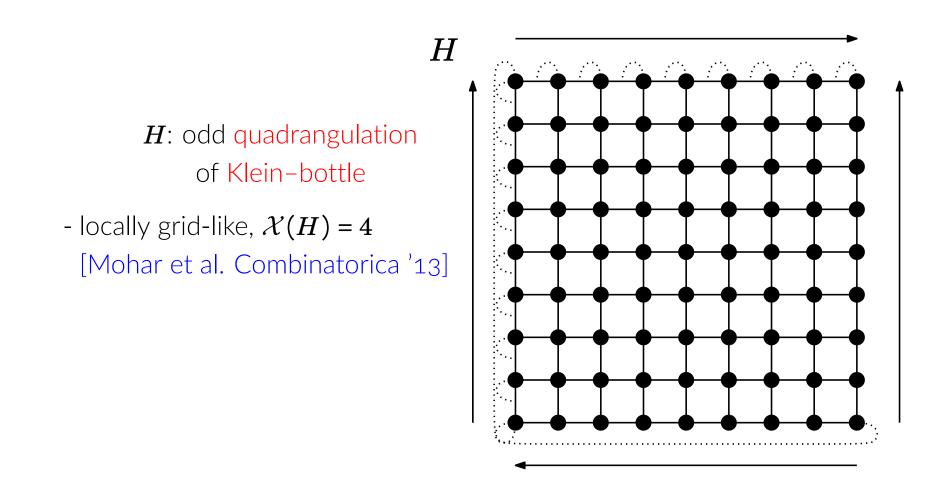


- Is it possible to "sandwich" quantum-LOCAL between weaker and stronger models?
 - yes! for some problems

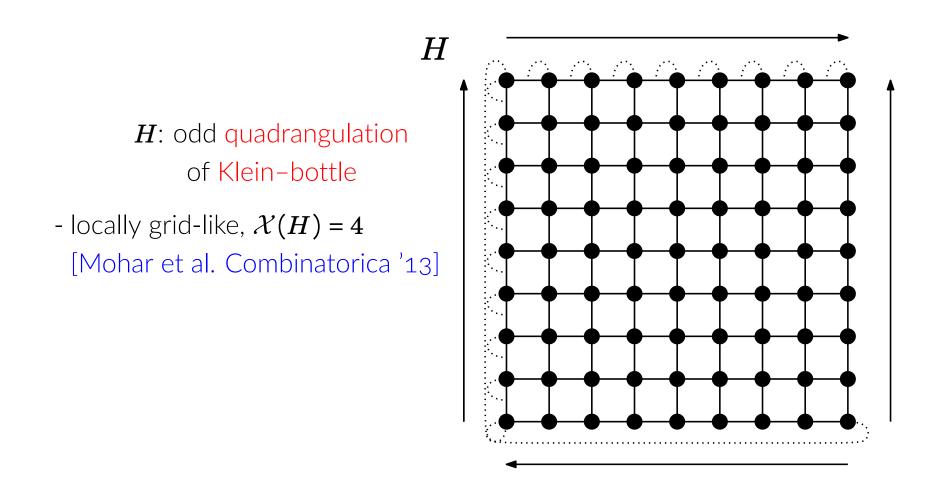
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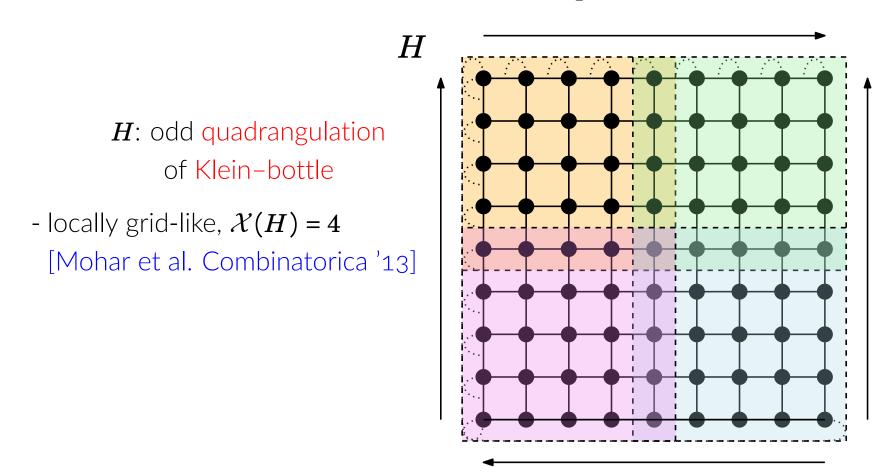
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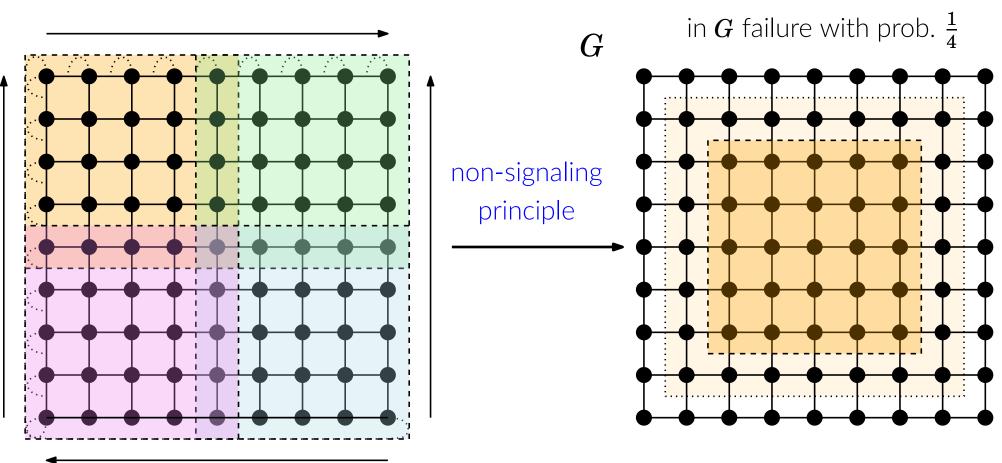
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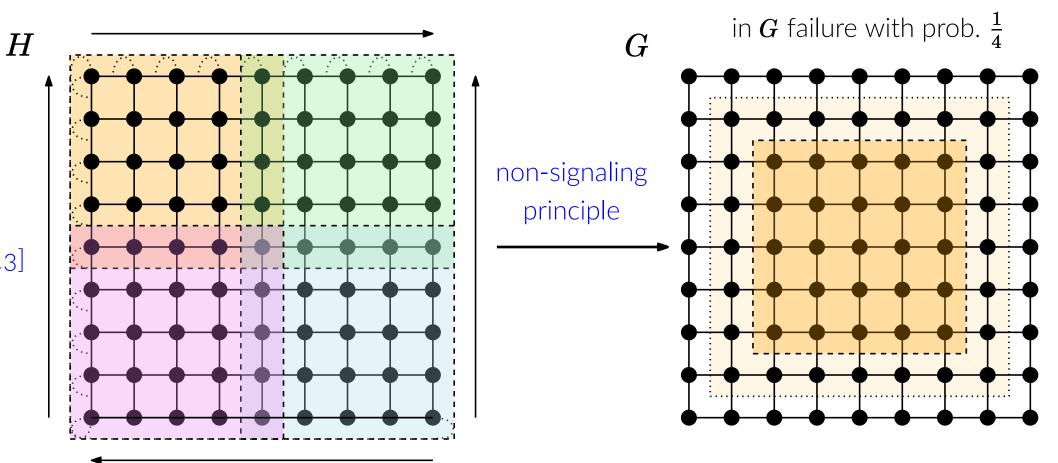


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• Boosting failure prob. is also possible



Graph-existential lower bound arguments based on indistinguishability

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What about other known lower bounds? E.g., 3-coloring cycles has complexity $\Theta(\log^* n)$ [Linial FOCS '87]

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 - no!
 - For any $\Theta(\log^* n)$ LCL Π on bounded degree graphs, there is a bounded-dependent distribution (T = O(1)) solving Π [Akbari et al. STOC '24]

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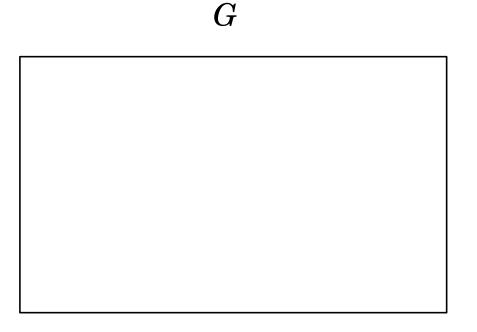
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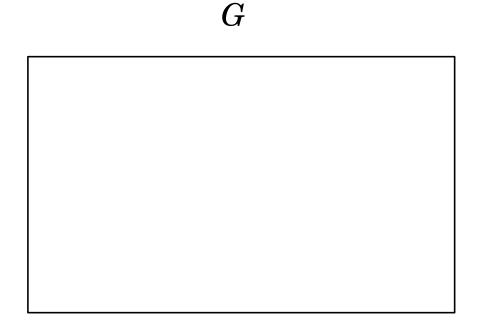
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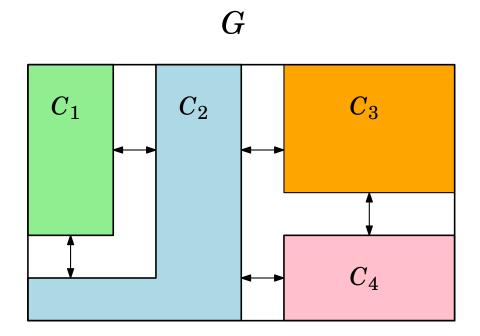
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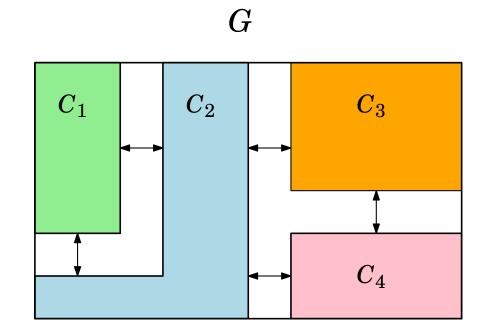
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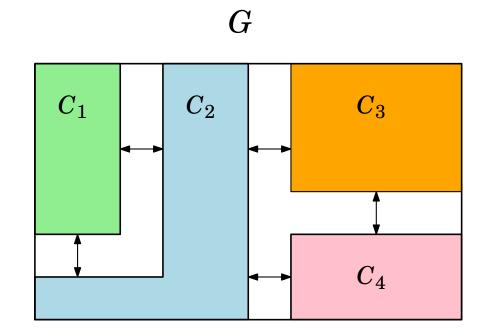
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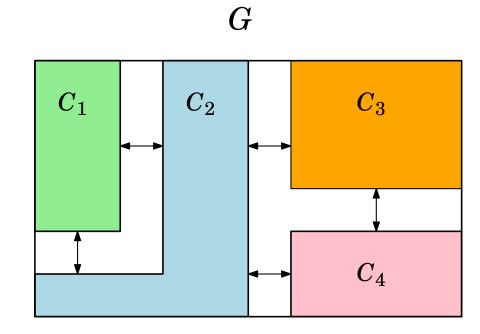
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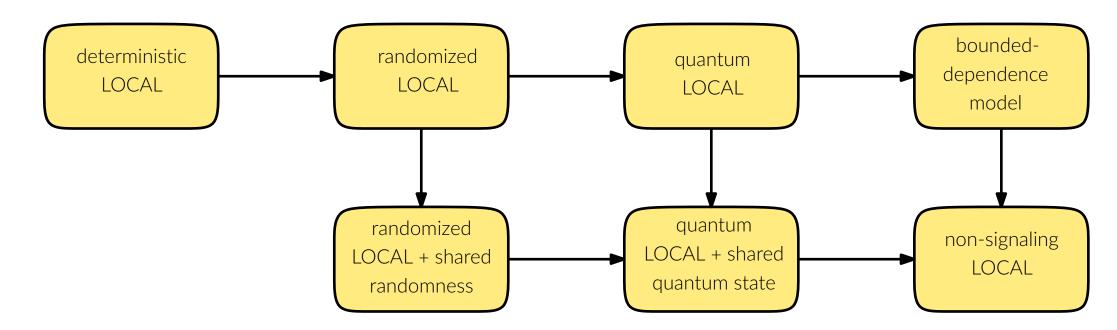
- ullet Bounded dependent distribution with locality $T \Longrightarrow \mathsf{LOCAL}$ algorithm with locality $ilde{O}(\sqrt{nT})$
- LOCAL complexity is $\Theta(n) \Longrightarrow$ bounded-dependence complexity $\tilde{\Omega}(n)$ (same for quantum-LOCAL)
- bounded-dependence (or quantum-LOCAL) complexity $O(1) \Longrightarrow \text{LOCAL}$ complexity is $\tilde{O}(\sqrt{n})$

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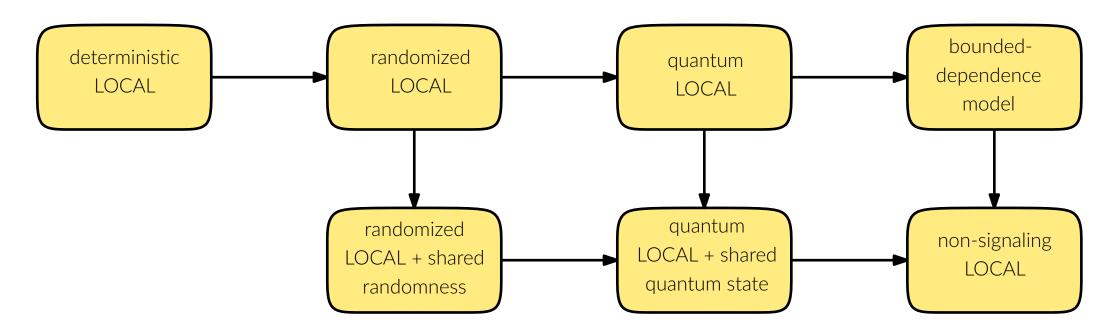
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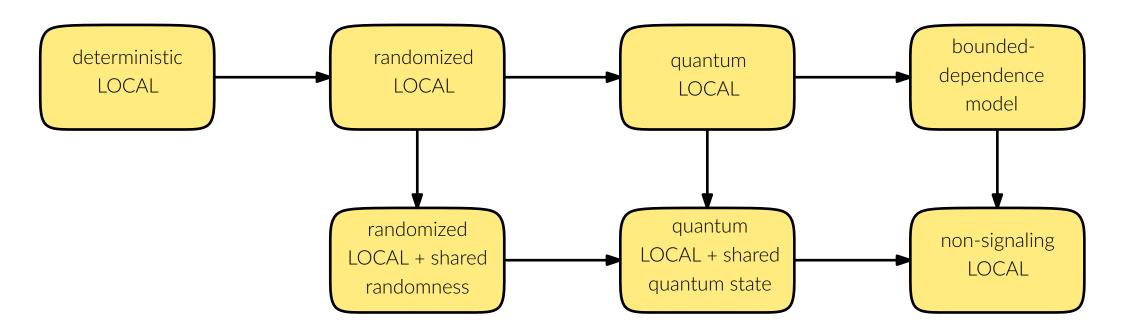
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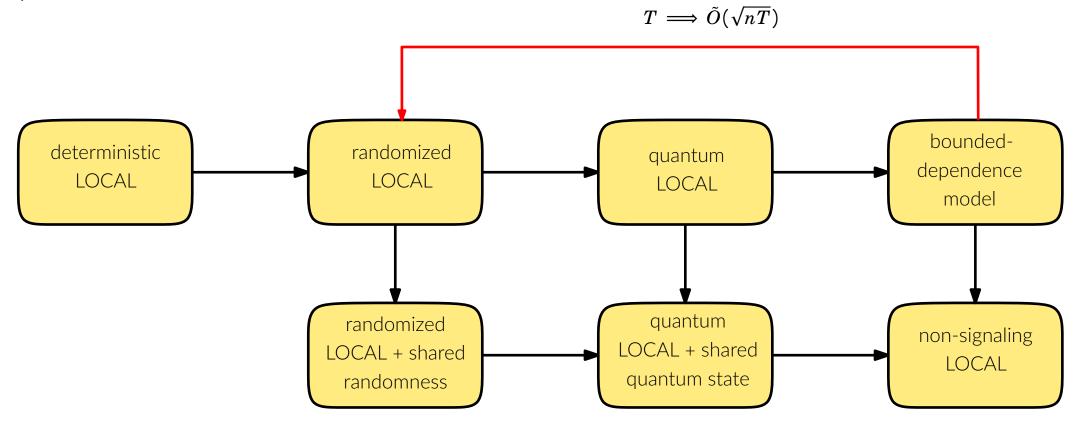
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