

Network Constructors: A Model for Programmable Matter

Paul G. Spirakis

joint work with

Othon Michail

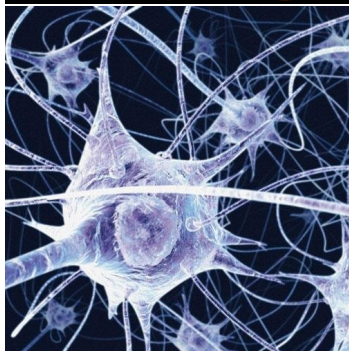
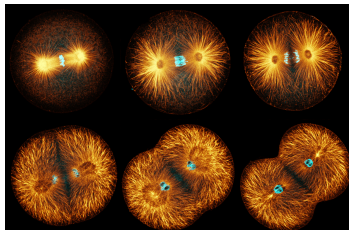
Department of Computer Science, University of Liverpool, UK
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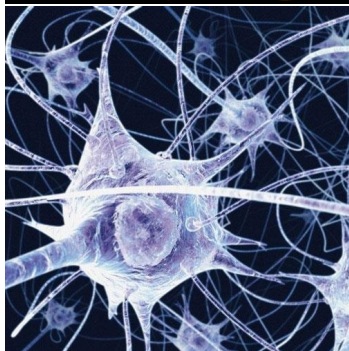
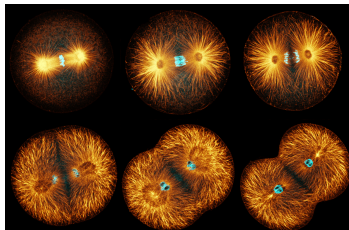
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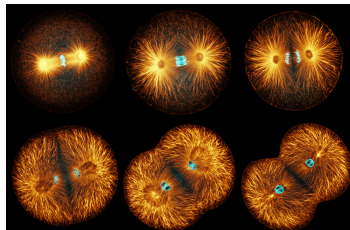
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- Usually collections of very large numbers of **simple distributed entities**
- Higher-level properties are the outcome of coexistence and constant interaction (cooperative and/or competing) of such entities
- Goal:
 - **Reveal the algorithmic aspects of physical systems**
 - **Develop innovative artificial systems inspired by them**



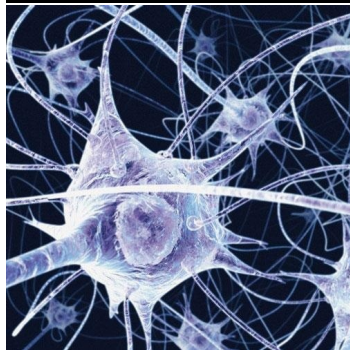
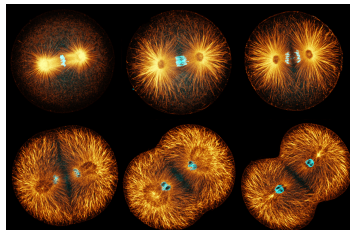
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- **Population Protocols** [AADFP, PODC '04] are formally equivalent to Chemical Reaction Networks [Doty, SODA '14]
- **Network Constructors** [Michail, Spirakis, PODC '14; Michail, PODC '15]: abstract and simple model of distributed network formation
- **Algorithmic self-assembly of DNA**: DNA tiles binding to other tiles via Watson-Crick complementary sticky ends
- Models of programmable matter equipped with active mobility/actuation mechanisms

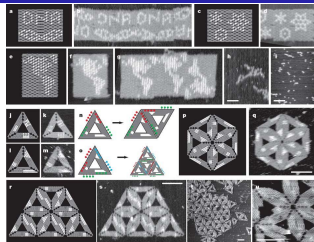
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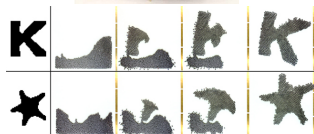
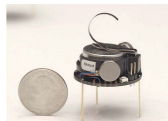
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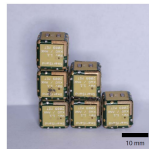
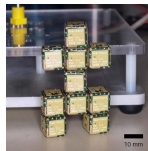
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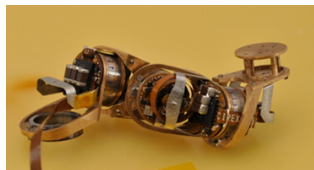
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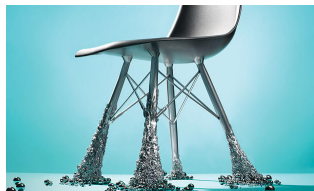
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- Incorporation of information to the **physical world**
- Plausible future outcome of progress in high-volume nanoscale assembly
- Physical realization of any computer-generated object
- Profound implications for how we think about **chemistry** and **materials**
- **Materials** will become user-programmed, smart, and adaptive
- It will change the way we think about **engineering** and **manufacturing**

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[Angluin, Aspnes, Diamadi, Fischer, and Peralta, PODC '04]

Distributed model. n computational entities, called **nodes**

- 1 interact in **pairs**
- 2 cannot control their interactions
 - passive mobility, like particles in a well-mixed solution
 - fair adversary or uniform random scheduler
- 3 have constant memory (uniformity)
- 4 **do not** have unique ids (anonymity)
- 5 $\delta : Q \times Q \rightarrow Q \times Q$: **transition function**

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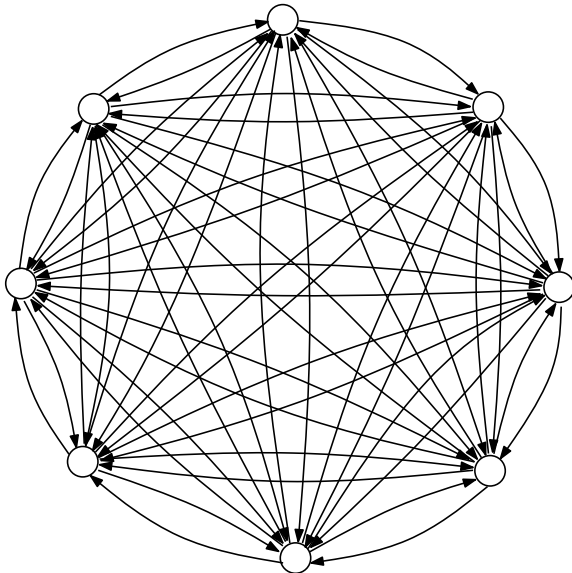
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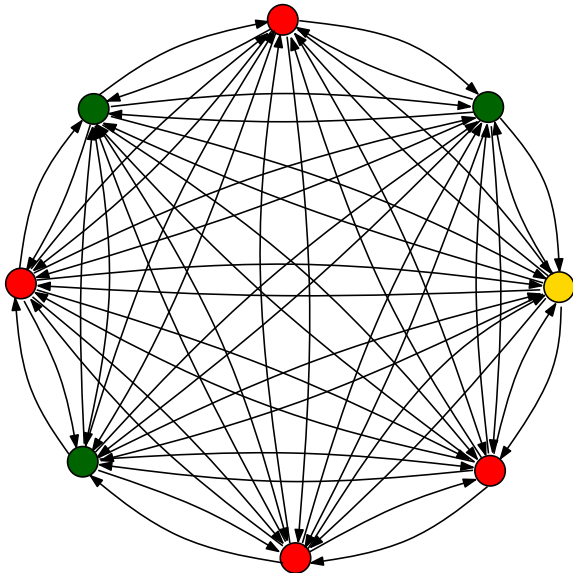
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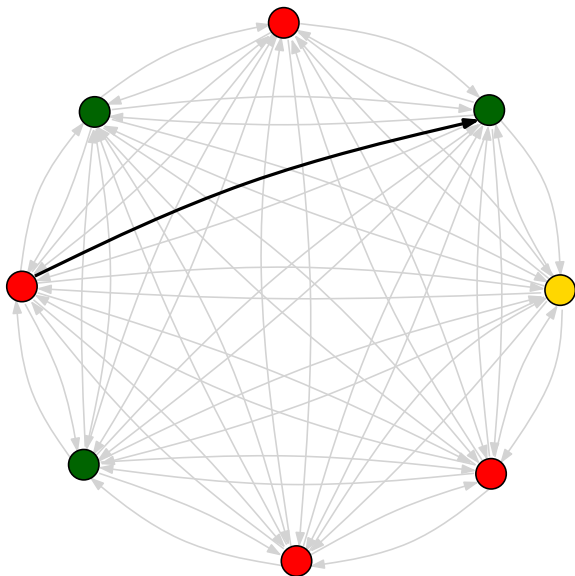
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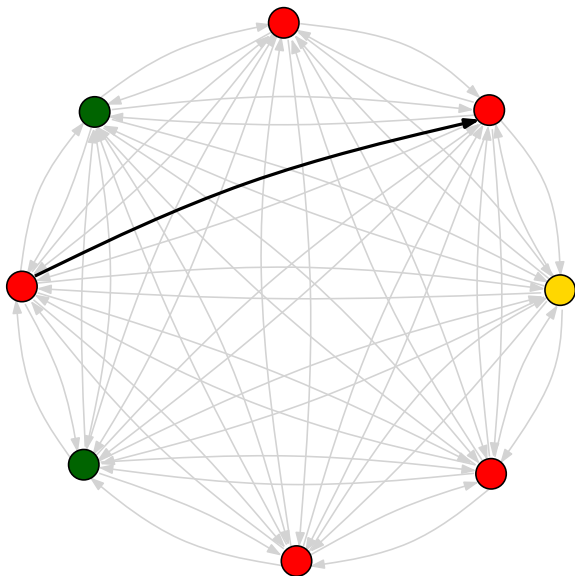
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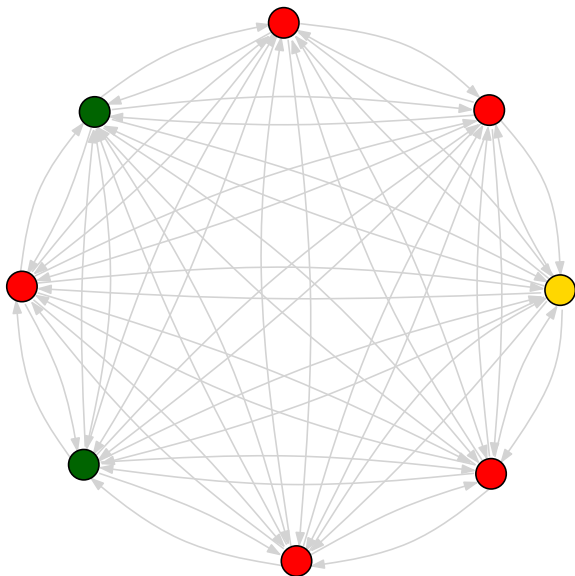
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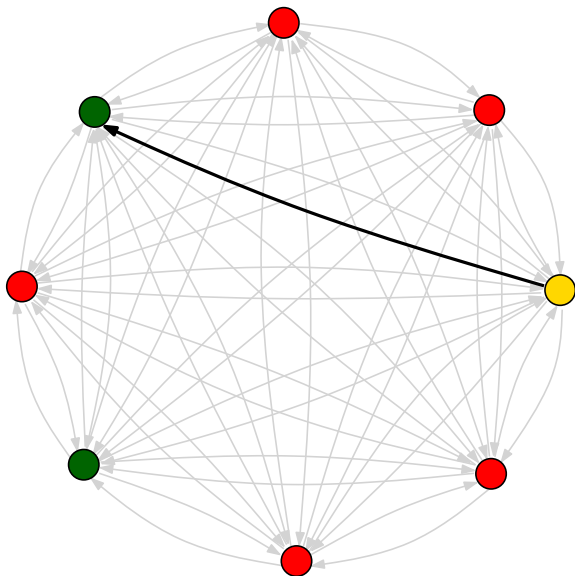












- May be viewed as an abstraction of “fast-mixing” physical systems
- Chemical Reaction Networks
 - Finite sets of chemical reactions such as $A + B \rightarrow A + C$
 - Promising programming language for molecular control circuitry
 - Some CRNs can simulate space-bounded Turing machines
- PPs are formally equivalent to CRNs [Doty '14]
 - Consequence: bounds for PPs, usually translate to inherent properties of natural systems
 - e.g., time-consuming to generate exact quantities of molecular species quickly [DS15]
- Relations to self-assembly, gene regulatory networks, opinion spreading, and antagonism of species models
- A predicate is computable by PPs iff it is semilinear [AAER '07]

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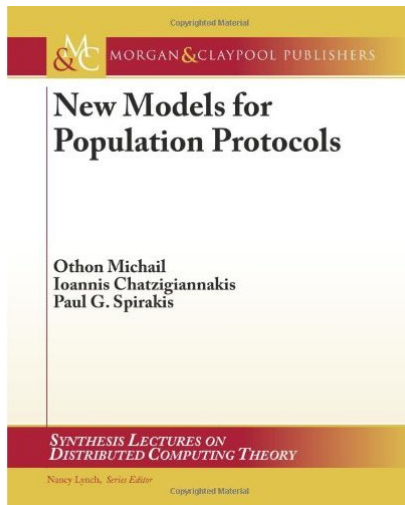
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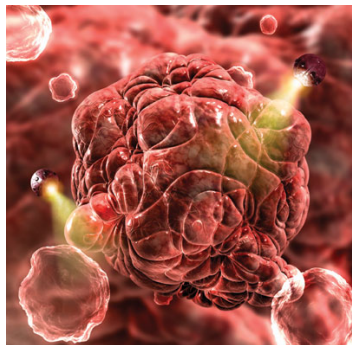
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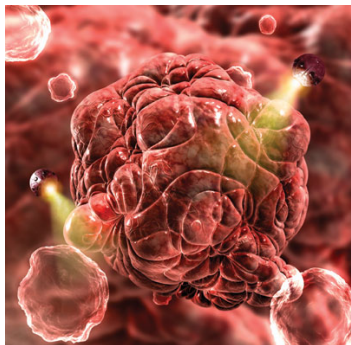


- n tiny computational devices
- Injected into a human circulatory system for monitoring/treating
- Move and interact passively (blood flow)
- Cooperation: can create bonds with each other
- Self-assemble into a desired global structure/network
- The artificial population evolves greater complexity, better storage capacity, and adapts and optimizes its performance to the specific task to be accomplished

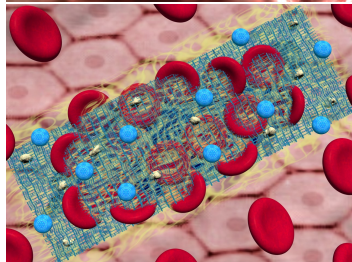
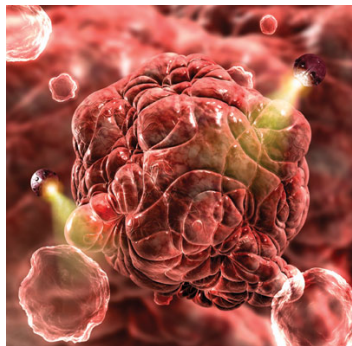
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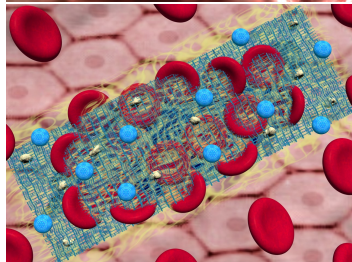
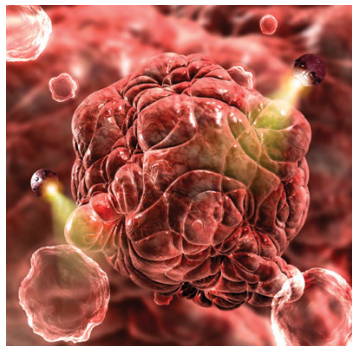
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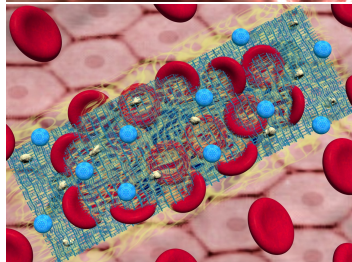
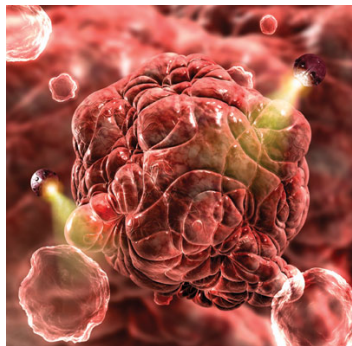
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[Michail and Spirakis, PODC '14 and Distrib. Comput. '16]

Fundamental problem: Algorithmic distributed construction of an actual communication topology

- Processes can form/delete connections between them
- Physical or virtual connections depending on the application
- on/off case: a connection either exists (active) or does not exist (inactive)
- Initially all connections are inactive
- Goal: End up with a desired stable network

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Our Goal: Determine what **stable structures** can result in such systems, **how fast**, and **under what conditions** (e.g. by what underlying codes/ reaction-rules)

- Principles of algorithmic network formation
- Programmable matter
- Reconfigurable robotics
- Medicine (diagnosis/treatment)
- Smart material (that self-built, self-adjust, ...)
- Biomaterial manufacture
- Modeling and understanding network formation by physical/chemical/biological processes (e.g. molecules reacting in a well-mixed solution)

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- 1 Q : finite set of **node-states**,
 - 2 $q_0 \in Q$: **initial node-state**,
 - 3 $Q_{out} \subseteq Q$: set of **output node-states**, and
 - 4 $\delta : Q \times Q \times \{0, 1\} \rightarrow Q \times Q \times \{0, 1\}$: the **transition function**
- In every step, a **pair uv** is selected by the scheduler and u, v interact according to δ
 - **Fair scheduler**: If a configuration is reachable infinitely often then it is eventually reached
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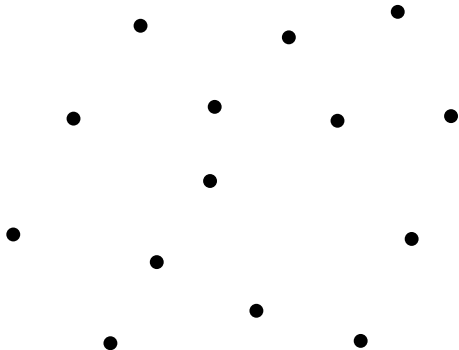
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- 2 states: black and red
- Initially all black
- Construct a global star
- Program:

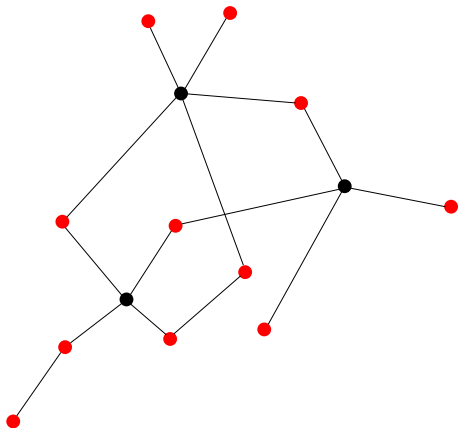
$$(b, b, 0) \rightarrow (b, r, 1)$$

$$(r, r, 1) \rightarrow (r, r, 0)$$

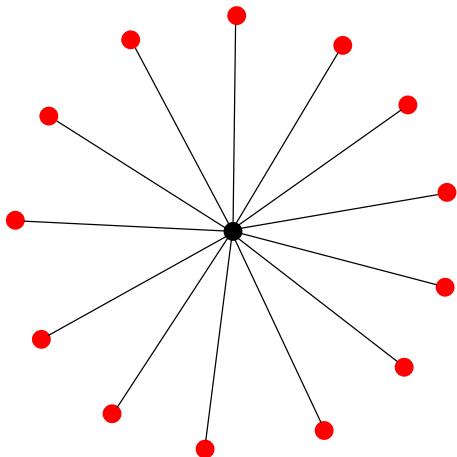
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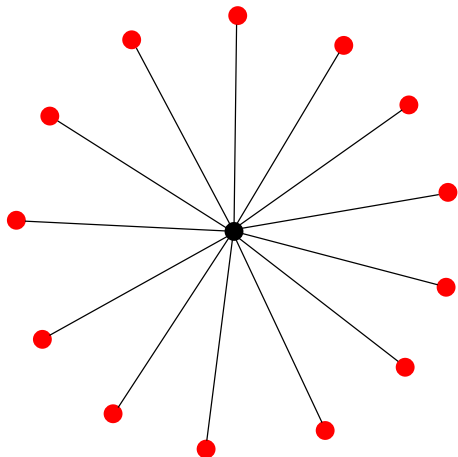
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- Size: 2 states
- Time: $O(n^2 \log n)$
- Optimal w.r.t. both
- Time = # interactions (sequential)
 - Parallel time is on the average $\frac{1}{n}$ sequential time



Global Line: For all n , the n processes must construct a **spanning line**

Theorem (Generic Lower Bound)

*The expected time to convergence (always under the uniform random scheduler) of any protocol that constructs a **spanning network** is $\Omega(n \log n)$.*

Proof.

Consider the time at which the last edge is activated. By that time, all nodes have some active edge incident to them, thus every node has interacted at least once. The latter can be shown to require an expected number of $\Theta(n \log n)$ steps. □

Theorem (Line Lower Bound)

*The expected time to convergence of any protocol that constructs a **spanning line** is $\Omega(n^2)$.*

- Protocol *Simple-Global-Line*:

$$(q_0, q_0, 0) \rightarrow (q_1, l, 1)$$

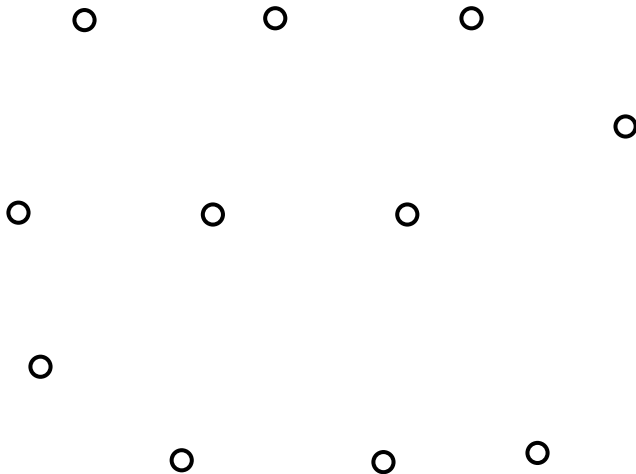
$$(l, q_0, 0) \rightarrow (q_2, l, 1)$$

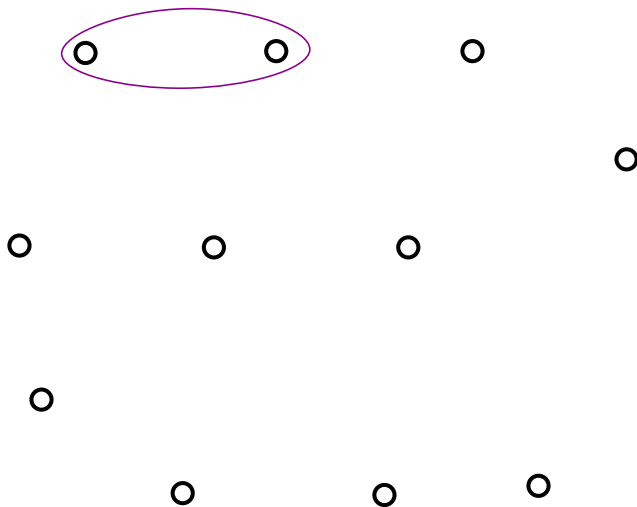
$$(l, l, 0) \rightarrow (q_2, w, 1)$$

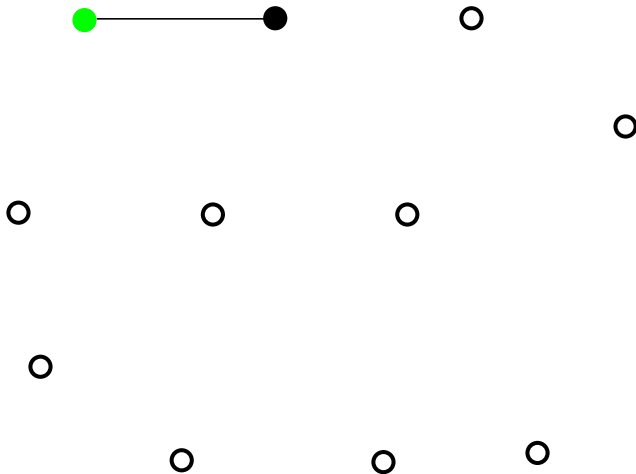
$$(w, q_2, 1) \rightarrow (q_2, w, 1)$$

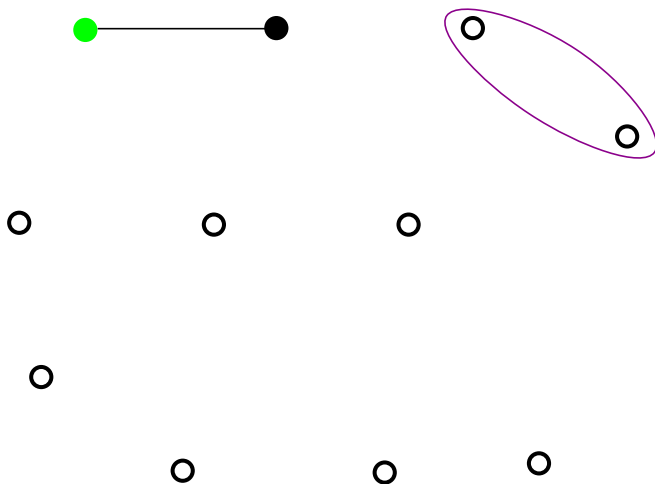
$$(w, q_1, 1) \rightarrow (q_2, l, 1)$$

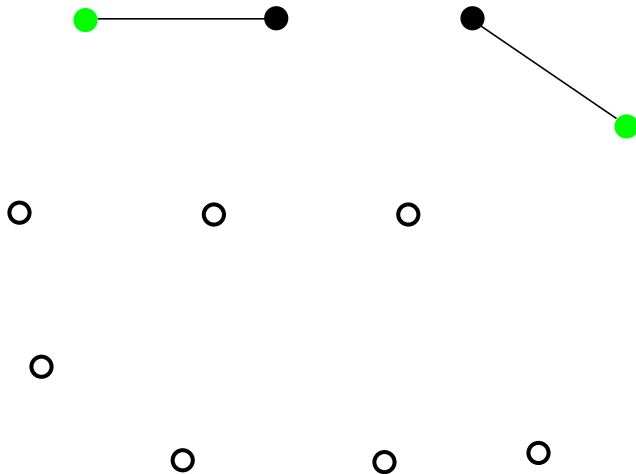
- Every node remembers its degree (q_0, q_1, q_2)
- Every line has a unique leader (endpoint: l , internal: w)
- Lines expand towards isolated nodes and merge to other lines via l
- After merging, w performs a random walk to reach an endpoint

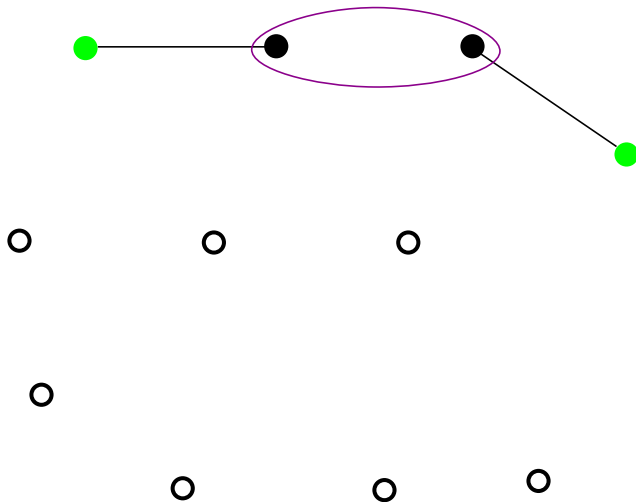


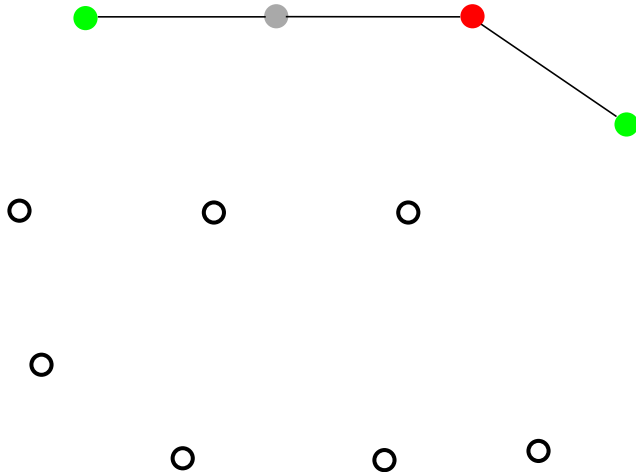


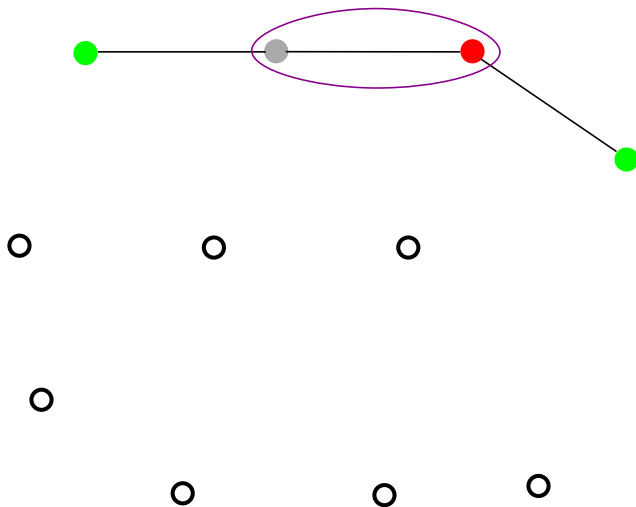


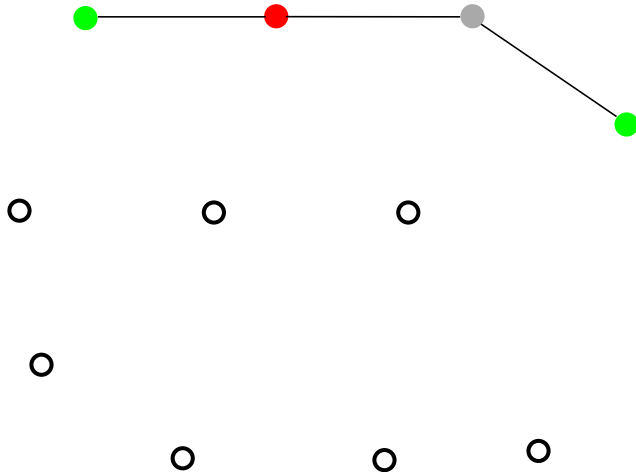


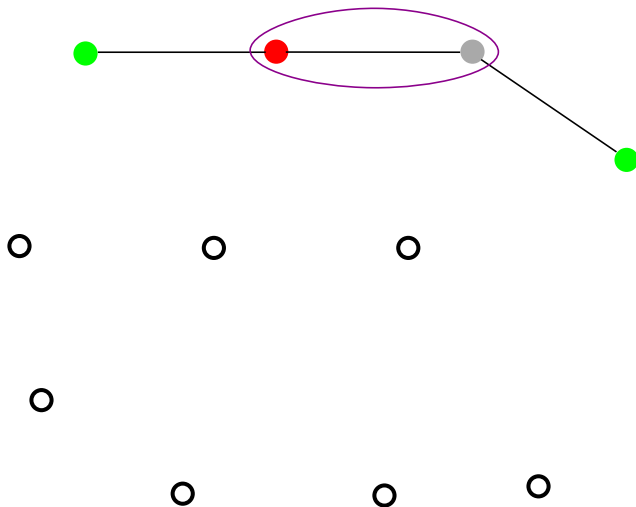


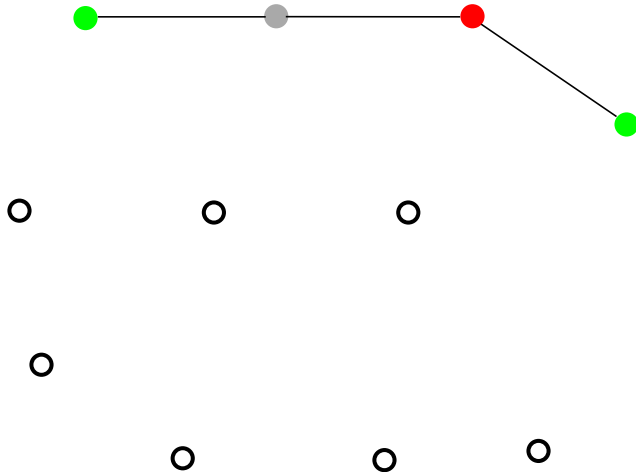


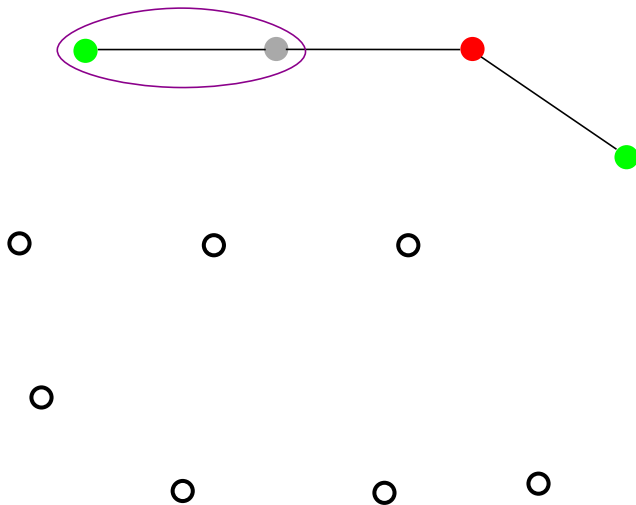


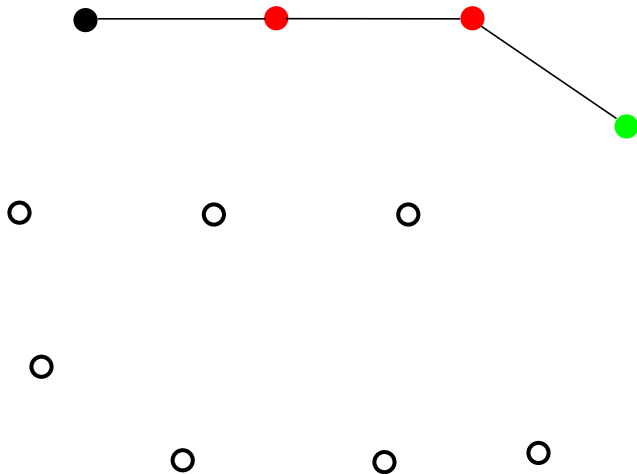


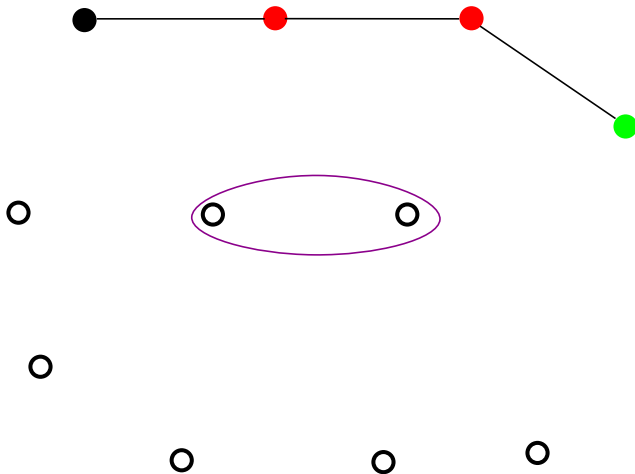


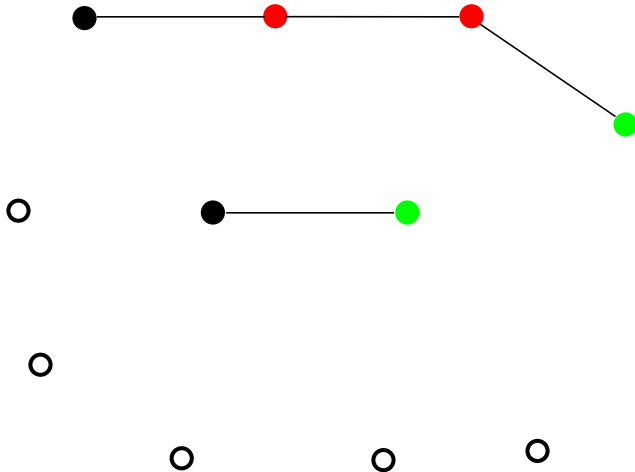


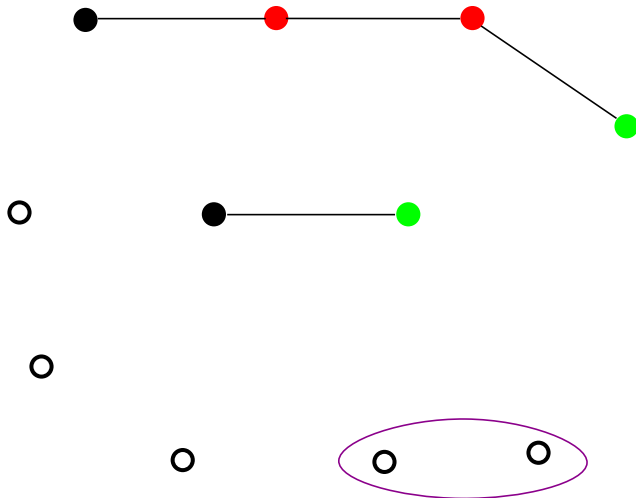


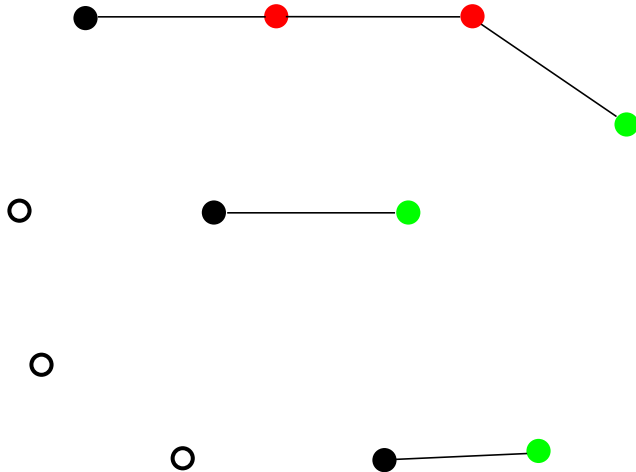


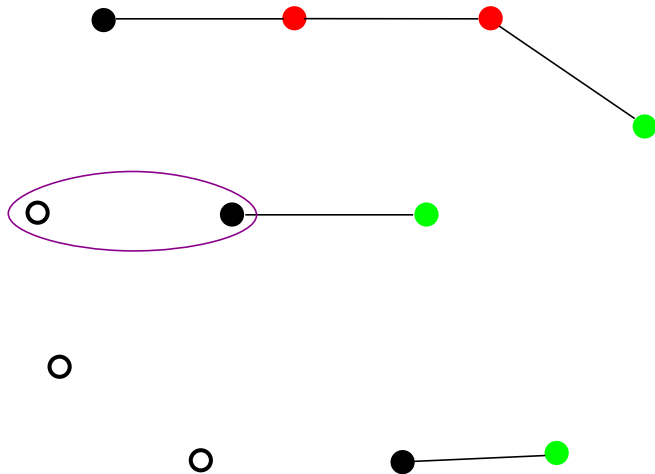


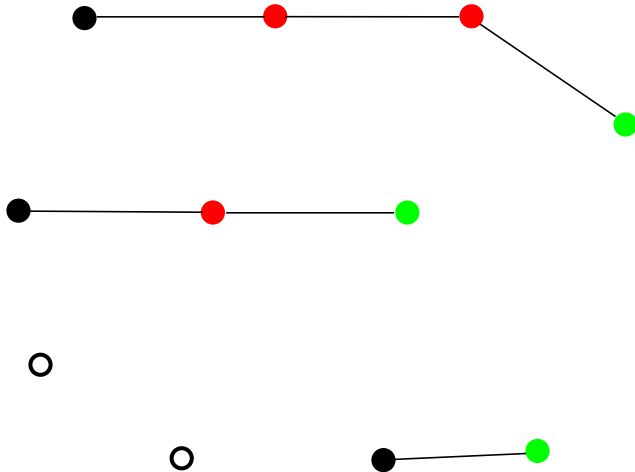


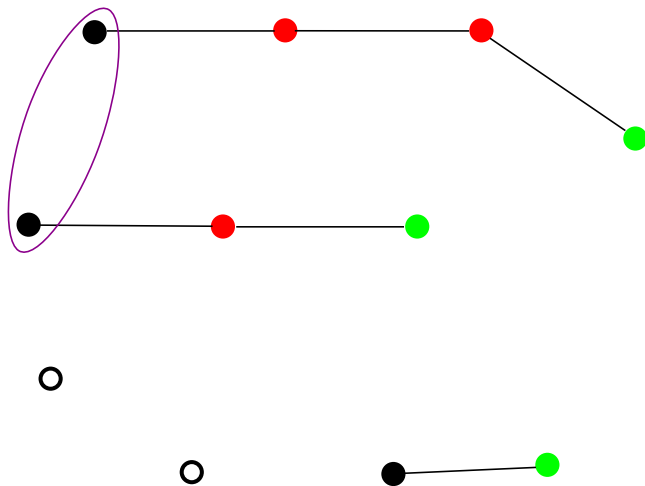


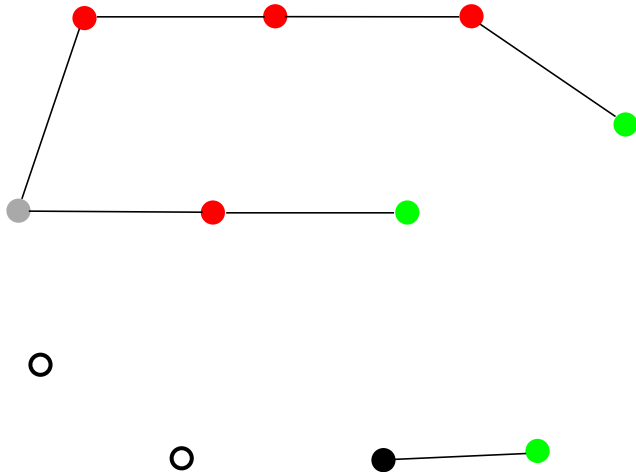


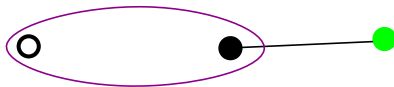
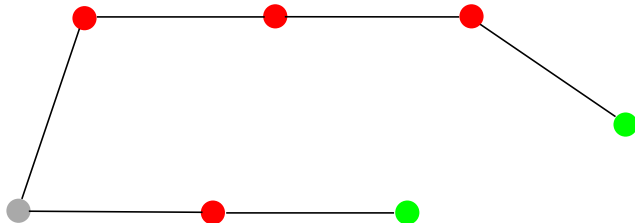


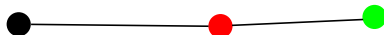
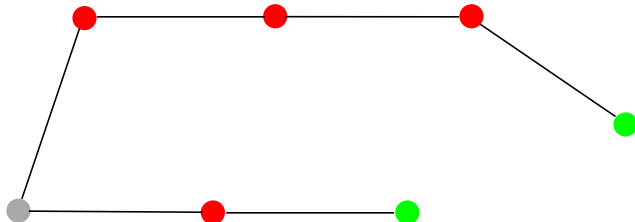


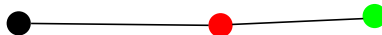
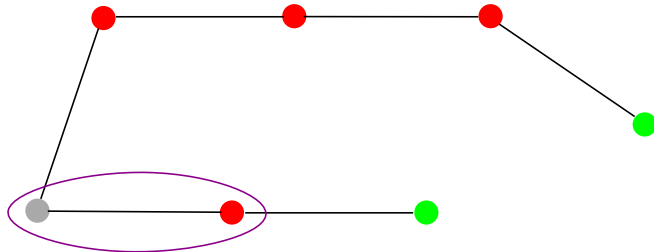


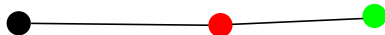
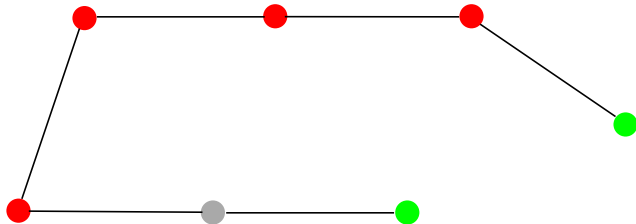


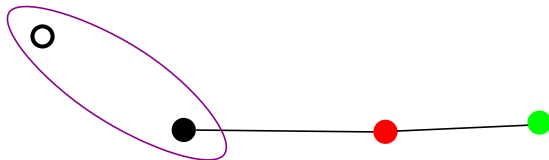
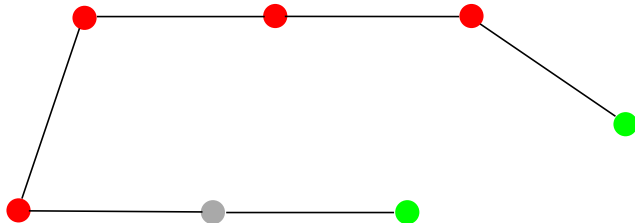


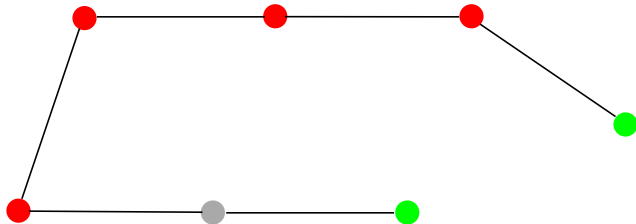


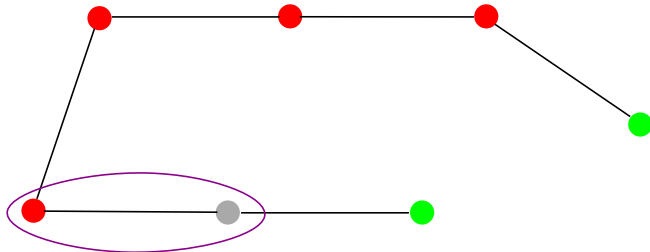


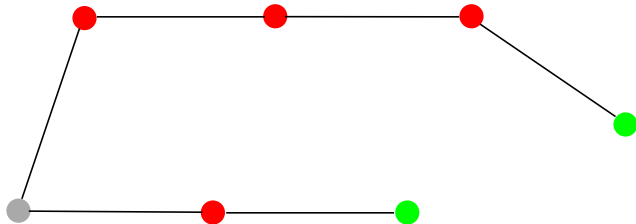


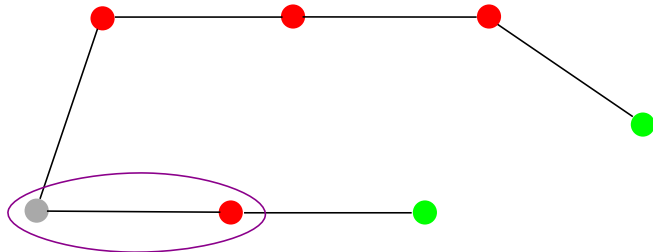


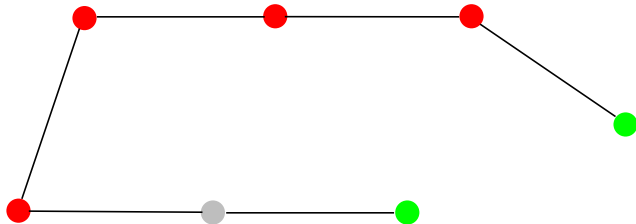


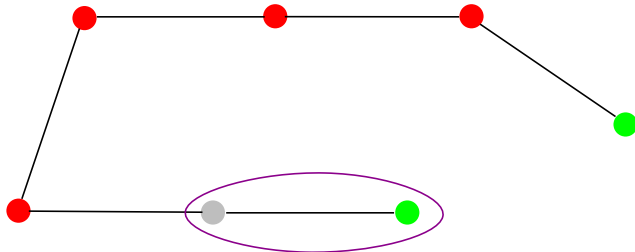


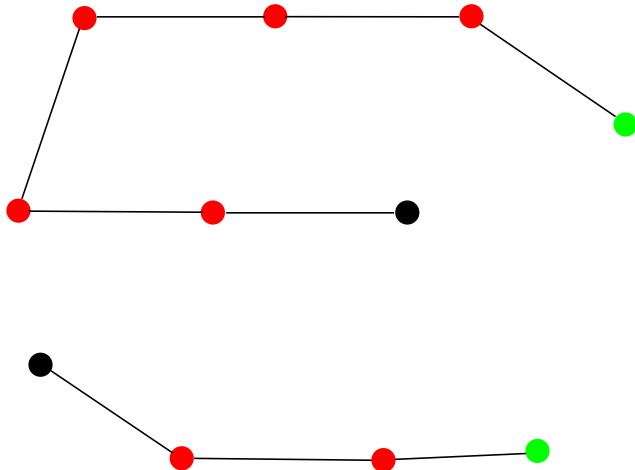


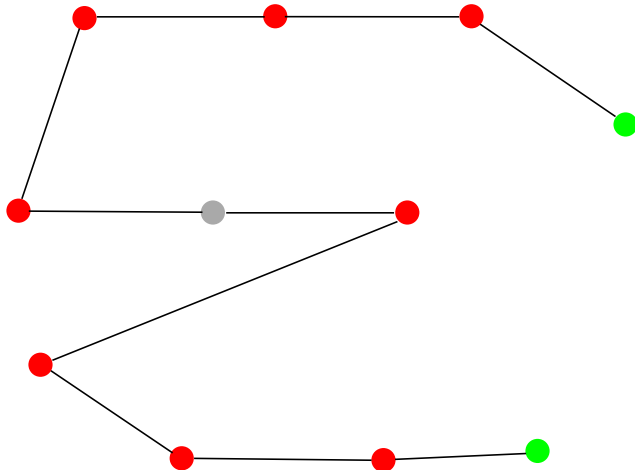


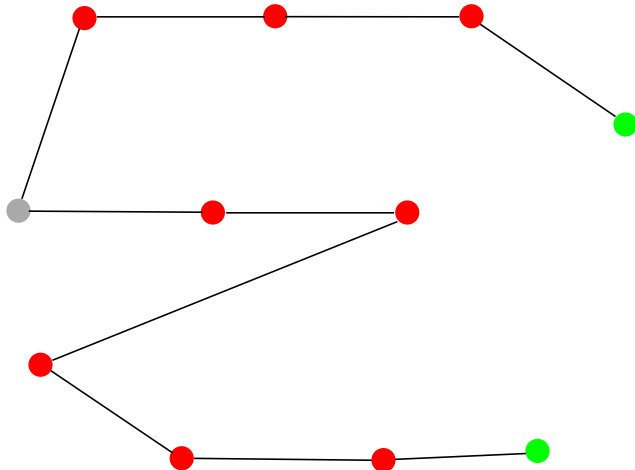


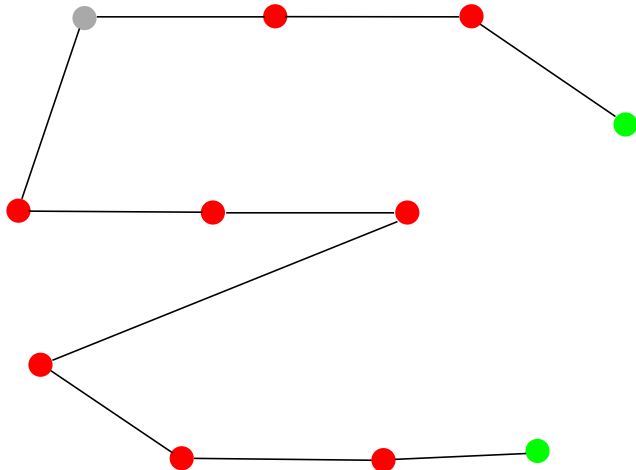


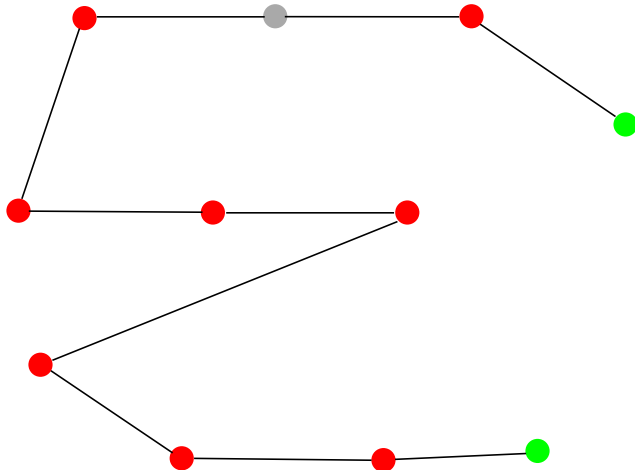


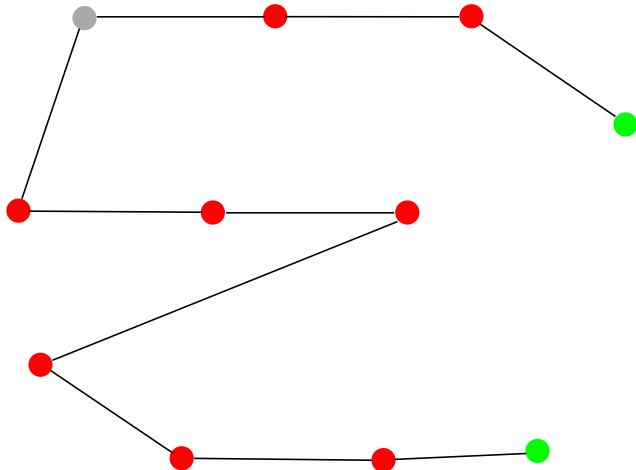


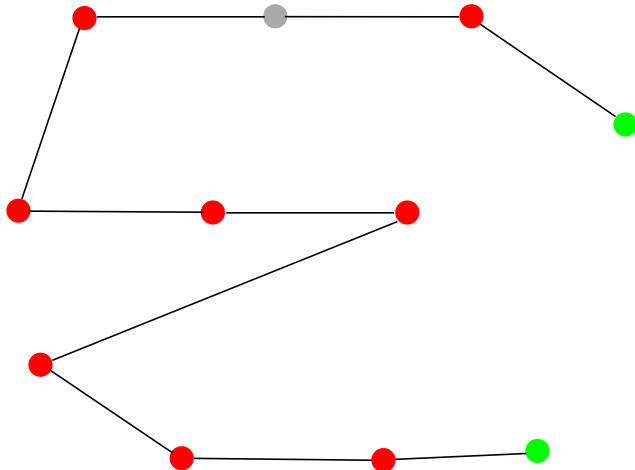


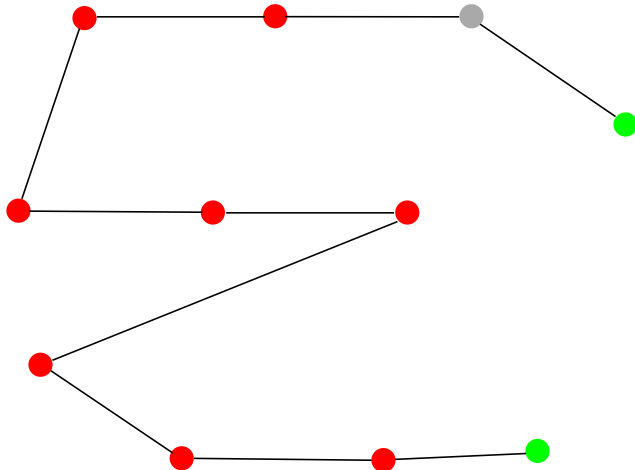


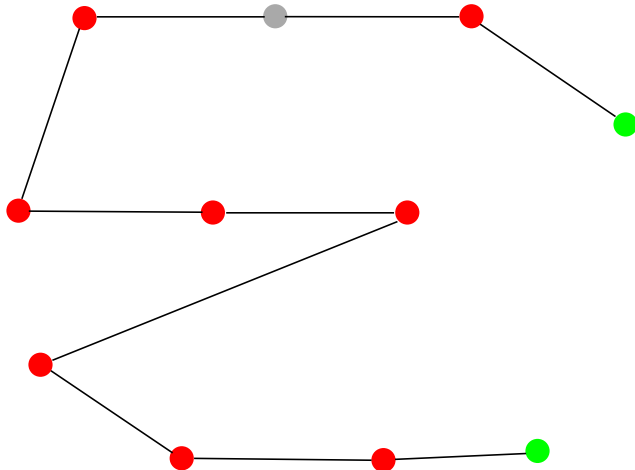


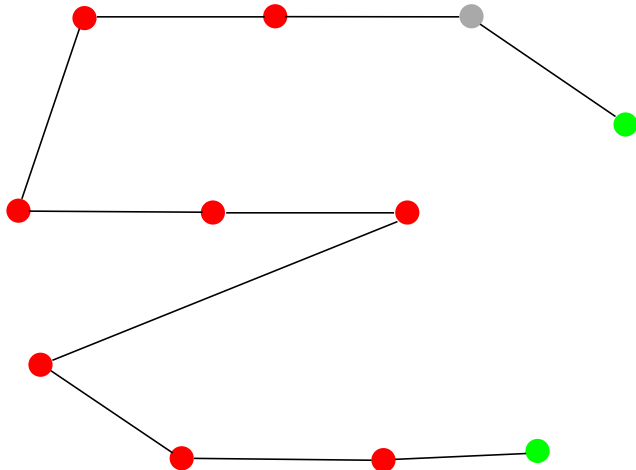




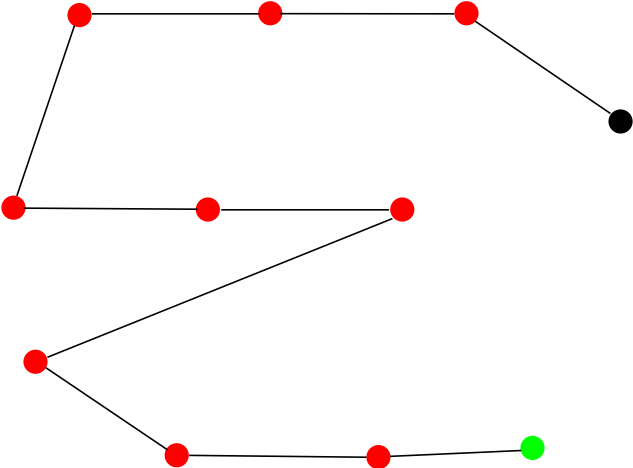








Simple-Global-Line Protocol



Theorem

Protocol Simple-Global-Line constructs a spanning line. It uses 5 states and its expected running time is $\Omega(n^4)$ and $O(n^5)$.

Proof

We have to prove two things:

- There is a set \mathcal{S} of output-stable configurations whose active network is a spanning line*
- For every reachable configuration C it holds that $C \rightsquigarrow C_s$ for some $C_s \in \mathcal{S}$*
- Spanning line, non-leader endpoints in state q_1 , non-leader internal nodes in q_2 , and unique leader either in l (endpoint) or in w (internal)*
- Any reachable C is a collection of active lines with unique leaders and isolated nodes. It is not hard to present a finite sequence of transitions that converts C to a $C_s \in \mathcal{S}$.*

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- 1 There is a *set \mathcal{S} of output-stable configurations* whose active network is a *spanning line*
 - 2 For every reachable configuration C it holds that $C \rightsquigarrow C_s$ for some $C_s \in \mathcal{S}$
- 1 *Spanning line, non-leader endpoints in state q_1 , non-leader internal nodes in q_2 , and unique leader either in l (endpoint) or in w (internal)*
 - 2 *Any reachable C is a collection of active lines with unique leaders and isolated nodes. It is not hard to present a finite sequence of transitions that converts C to a $C_s \in \mathcal{S}$.*

Theorem

Protocol Simple-Global-Line constructs a spanning line. It uses 5 states and its expected running time is $\Omega(n^4)$ and $O(n^5)$.

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- **Upper bound:** $(n - 2)[O(n^2) + O(n^4)] = O(n^5)$
- **Lower Bound:**
 - *w.h.p. constructs $\Theta(n)$ different lines of length 1 during its course*
 - *$k = \Theta(n)$ disjoint lines $\Rightarrow k - 1 = \Theta(n)$ distinct merging processes*
 - *Let t_k be the first time at which there is a line of length $\geq k/2$.*
 - *It holds that $k/4 \leq t_k \leq k/2 + 1$.*
 - *Therefore, there are at least $t_k - t_{k-1} \geq (k - 1)(1/2 - 1/4) = k/4 - 1/2$ merging processes that merge lines of length $\geq k/4$ into lines of length $\geq k/2$.*



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$$(l, q_0, 0) \rightarrow (q_2, l, 1)$$

$$(l, l, 0) \rightarrow (q'_2, l', 1)$$

$$(l', q_2, 1) \rightarrow (l'', f_1, 0)$$

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- **Avoid mergings** as they seem to consume much time
 - even **deterministic merging** takes time $\Omega(n^3)$ and $O(n^4)$
- When leaders of lines interact, they play a **pairwise game**
- The **winner grows** by one towards the other line and **the loser sleeps**
- A **sleeping line** cannot increase any more and only **loses nodes** by lines that are still awake
- A single leader is guaranteed to always win and eventually remain unique and this occurs quite fast
- The leader makes progress (by one) in most interactions and every such progress is in turn quite fast
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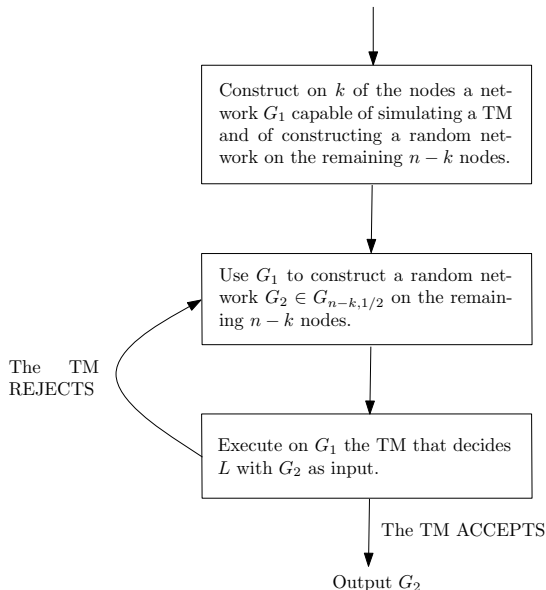
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Protocol	# states	Expected Time	Lower Bound
<i>Simple-Global-Line</i>	5	$\Omega(n^4)$ and $O(n^5)$	$\Omega(n^2)$
<i>Intermediate-Global-Line</i>	8	$\Omega(n^3)$ and $O(n^4)$	$\Omega(n^2)$
<i>Fast-Global-Line</i>	9	$O(n^3)$	$\Omega(n^2)$
<i>Cycle-Cover</i>	3	$\Theta(n^2)$ (optimal)	$\Omega(n^2)$
<i>Global-Star</i>	2 (optimal)	$\Theta(n^2 \log n)$ (optimal)	$\Omega(n^2 \log n)$
<i>Global-Ring</i>	9		$\Omega(n^2)$
<i>2RC</i>	6		$\Omega(n \log n)$
<i>kRC</i>	$2(k + 1)$		$\Omega(n \log n)$
<i>c-Cliques</i>	$5c - 3$		$\Omega(n \log n)$
<i>Graph-Replication</i>	12	$\Theta(n^4 \log n)$	

Question: Is there a generic constructor capable of constructing a large class of networks?

- 1 constructors that simulate a Turing Machine
- 2 a constructor that simulates a distributed system with names and logarithmic local memories
 - l : the binary length of the input of a TM
 - In what follows, $l = \Theta(n^2)$
 - **DGS**($f(l)$): the class of graph languages decidable by a TM of (binary) space $f(l)$ (input graph in adjacency matrix encoding)
 - **REL**($g(n)$): the class of graph languages constructible with useful space $g(n)$ (relation or on/off class)
 - **PREL**($g(n)$): (i) allow transitions that with probability $1/2$ give one outcome and with probability $1/2$ another (ii) all graphs must be constructed equiprobably

To construct a decidable graph-language L .

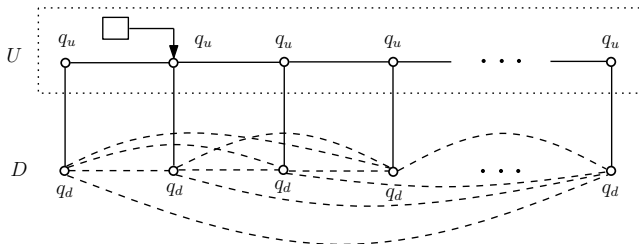


Theorem (Linear Waste-Half)

DGS($O(n)$) \subseteq **PREL**($\lfloor n/2 \rfloor$). In words, for every graph language L that is decidable by a $O(n)$ -space TM, there is a protocol that constructs L equiprobably with useful space $\lfloor n/2 \rfloor$.

Proof

- Partition the population into **equal sets U and D** and construct an active **perfect matching** between U and D
- Construct a **spanning line in U** (e.g. Fast-Global-Line)
- Organize the line into a **TM M**
- M must **compute a graph from L** and **construct it on D**
 - uniquely identify the nodes of D by their **distance from one endpoint**
 - to modify edge (i,j) mark appropriately the D -nodes at distances i and j from one endpoint



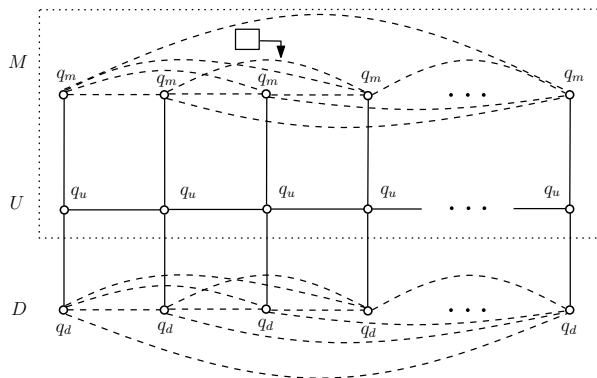
- M computes a graph G from L equiprobably: **activate/deactivate each edge of D equiprobably and independently of other edges**
- Then it simulates on input G the TM that decides L in \sqrt{I} space to determine whether $G \in L$
- The TM **rejects**: M repeats the random experiment to produce a new random graph G'
- The TM **accepts**: release the network, set D -nodes to q_{out}
- **Reinitialize** whenever the global line protocol makes progress □

- Nodes **cannot detect termination** and **cannot sequentially compose subroutines**
 - Instead, all subroutines must be **executed in parallel** and become **“sequentialized” by perpetual reinitializations**
- Executed whenever a line on U -nodes expands
- The protocol **“assumes”** that no further expansions will occur, **restores** the components of the simulation to their original values, ensures that each U -node has a D -neighbor, and starts drawing a new random graph
- The assumption may be wrong several times
- However, **eventually the assumption of the protocol will be correct**
 - The simulation will be reinitialized and executed for the last time on the correct sets U and D

Later we shall discuss **terminating protocols that are correct w.h.p.**

Theorem (Linear Waste-Two Thirds)

DGS($O(n^2) + O(n)$) \subseteq **PREL**($\lfloor n/3 \rfloor$). That is, for every graph language L that is decidable by a $O(n^2)$ -space TM, there is a protocol that constructs L equiprobably with useful space $\lfloor n/3 \rfloor$.



Theorem (Logarithmic Waste)

DGS($O(\log n)$) \subseteq **PREL**($n - \log n$).

Proof.

- Construct a line to **count n in binary**, i.e. to occupy **$\log n$ cells**
- Release the constructed counter and reinitialize the other (free) nodes
- So, there is a **logarithmic memory with a leader**, knowing a **good estimate of $n - \log n$**
- Construct a **random graph on the free nodes**:
 - For every free node, let it toss a coin on one after the other its edges to other free nodes
 - To know when to stop, count on the line up to $n - \log n$ (known) \square

- A population consisting of n nodes can be partitioned into k *supernodes* each consisting of $\log k$ nodes, for the largest such k
- The *internal structure* of each supernode is a *line*, thus it can be operated as a *TM of memory logarithmic* in the total number of supernodes
- This amount of storage is sufficient for the supernodes to obtain *unique names* and exploit their names and their internal storage to realize *nontrivial constructions*
- We are interested in the networks that can be constructed *at the supernode abstraction layer*

Theorem (Partitioning into Supernodes)

*For every network G that can be constructed by k nodes having **local memories** $\lceil \log k \rceil$ and **unique names** there is a NET that constructs G on $n = k \lceil \log k \rceil$ nodes.*

[Michail and Spirakis, TCS '16]

- Minimal strengthenings of NETs that can maximize computational power
 - Also gain termination
- Initial configuration: any connected graph spanning V
- Ability to detect small local degrees
 - e.g., a node can detect if its active degree is 0
 - we can now simulate any constructor that assumes an empty initial network
- Pre-elected leader or pairs of nodes able to tell whether they have a neighbor in common

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- Gives the **maximum computational power** that one can hope for in this family of models
- Can compute with termination any symmetric predicate computable by a TM of space $\Theta(n^2)$, and no more than this, i.e., it is an **exact characterization**
- Symmetricity can only be dropped by UIDs or other means of maintaining an ordering of the nodes' inputs
- This power is **maximal** because the distributed space of the system is $\Theta(n^2)$
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- The leader (center) starts connecting with the q_0 s converting them to p_0 s (peripherals)
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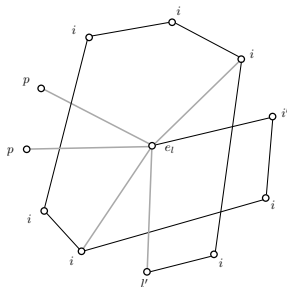
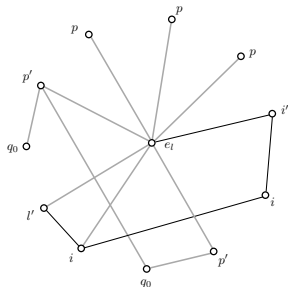
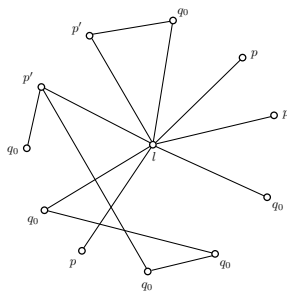
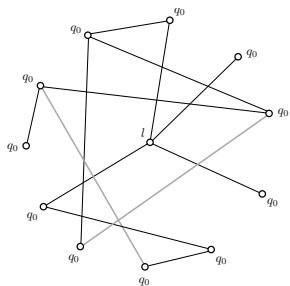
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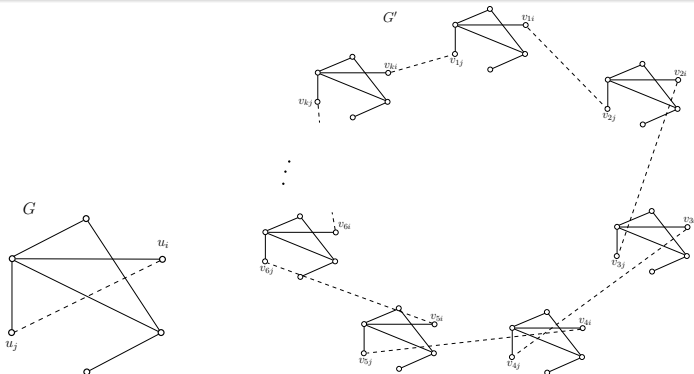
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An Example Execution



Theorem (Strong Impossibility)

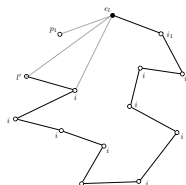
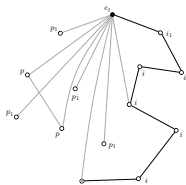
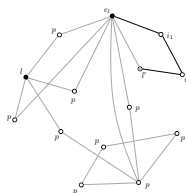
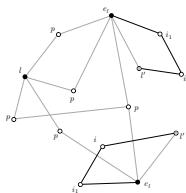
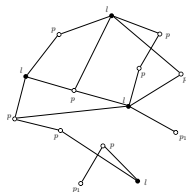
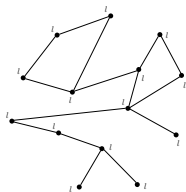
For every connected graph G with at least one cycle, there is an infinite family of graphs \mathcal{G} such that for every $G' \in \mathcal{G}$ every protocol (beginning from identical states on all nodes) that makes G acyclic may disconnect G' in some executions.



Theorem

By assuming that nodes are equipped with a *common neighbor detection mechanism* and have the ability to *detect local degrees 1 and 2*, *Protocol Line-Transformer* solves the *Terminating Line Transformation* problem in the setting in which all nodes are initially identical. Its running time is $O(n^3)$.

An Example Execution



Protocol	Leader	DD	CND	Expected Time	Lower Bound
Online-Cycle-Elimination	Yes	1	No	$\Theta(n^4)$	$\Omega(n^2 \log n)$
Line-Around-a-Star	Yes	1	No	$\Theta(n^2 \log n)$ (opt)	$\Omega(n^2 \log n)$
Line-Transformer	No	1,2	Yes	$O(n^3)$	$\Omega(n^2 \log n)$

- **Leader:** Whether it makes use of a pre-elected unique leader
- **DD:** what local degree detection is used
- **CND:** whether it uses common neighbor detection
- **Expected Time:** Expected running time under the uniform random scheduler
- **Lower Bound:** Best known lower bound

[Michail, PODC '15]

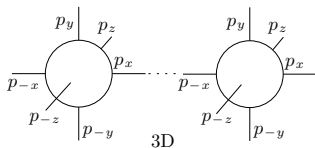
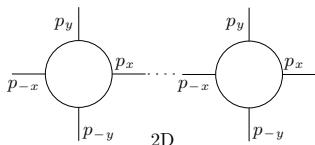
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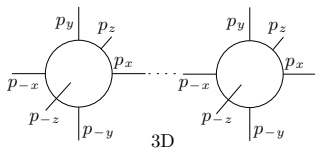
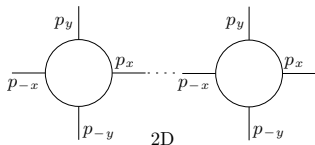
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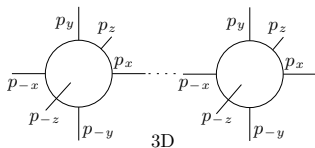
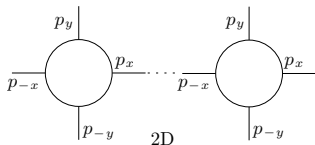
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- The **transition function** is now:

$$\delta : (Q \times P) \times (Q \times P) \times \{0, 1\} \rightarrow Q \times Q \times \{0, 1\},$$

$P = \{u, r, d, l\}$ is the **set of ports** in 2D

- In every step, a **pair** $(v_1, p_1)(v_2, p_2)$ is selected by the scheduler and v_1, v_2 interact via their p_1, p_2 ports according to δ
- **Valid configuration**: its connected components are subnetworks of the 2D grid network
- **Uniform random scheduler**: selects independently and uniformly at random from the permitted interactions (leading to valid config.)
- **Output shape**: nodes that are in output (or halting) states and edges between them that are active

- δ :

$$(L_u, u), (q_0, d), 0 \rightarrow (q_1, L_r, 1)$$

$$(L_r, r), (q_0, l), 0 \rightarrow (q_1, L_d, 1)$$

$$(L_d, d), (q_0, u), 0 \rightarrow (q_1, L_l, 1)$$

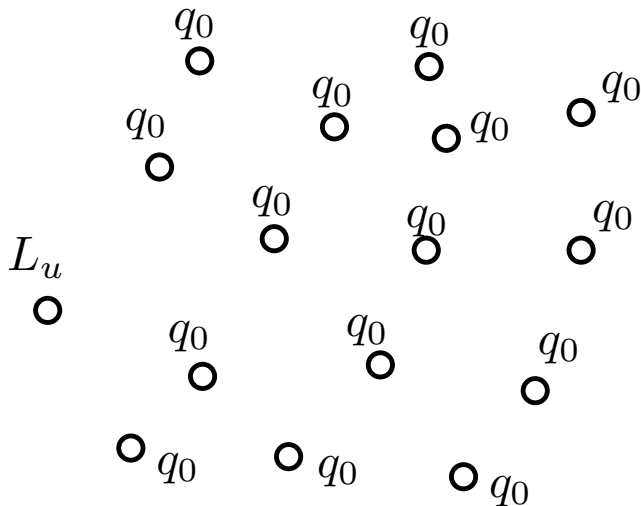
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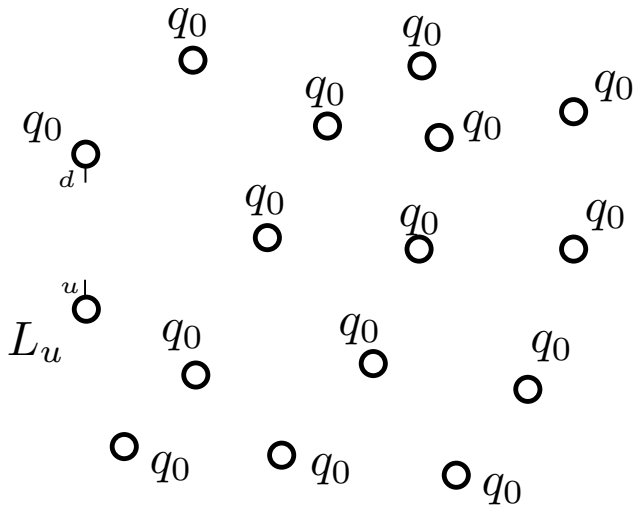
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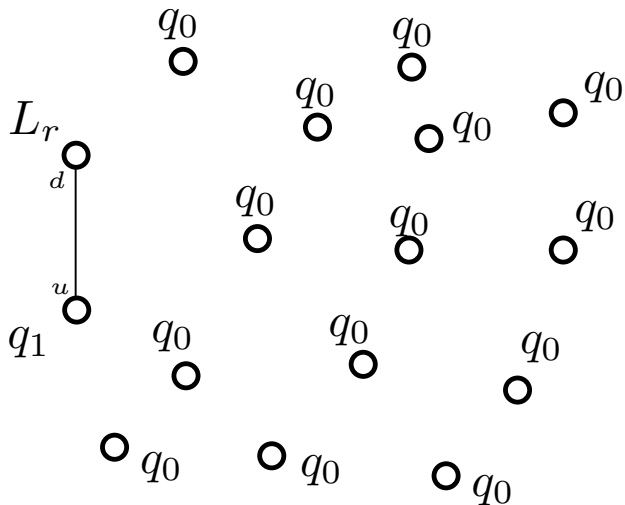
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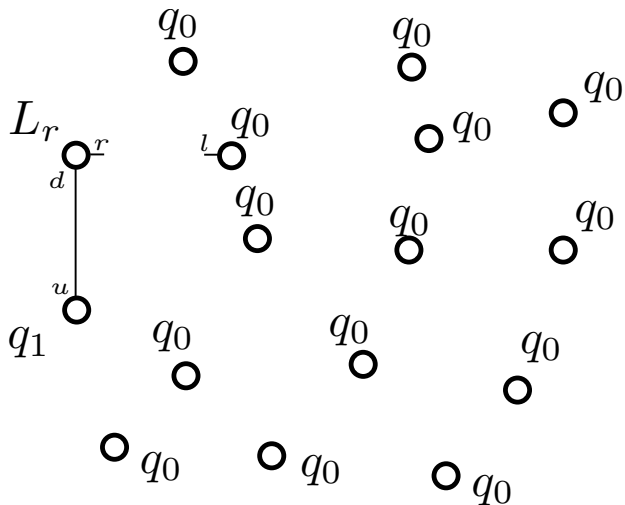
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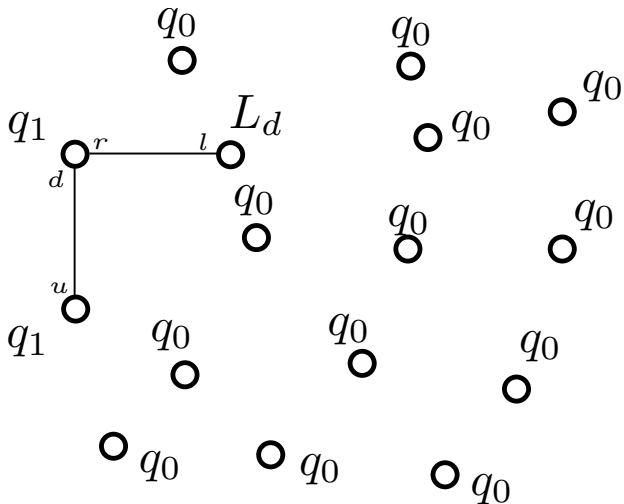
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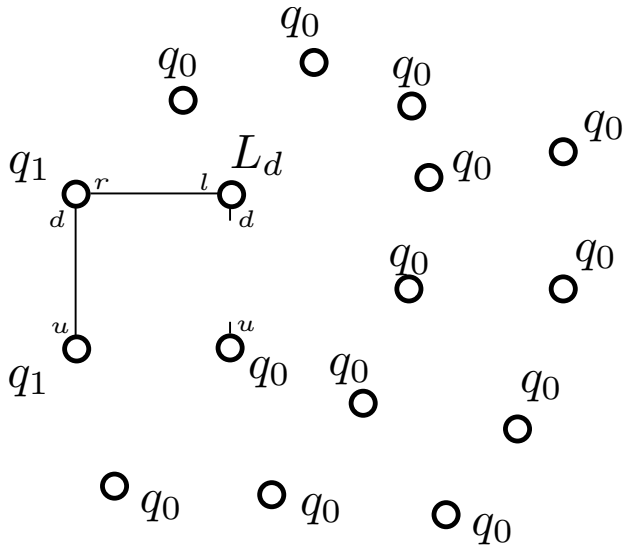


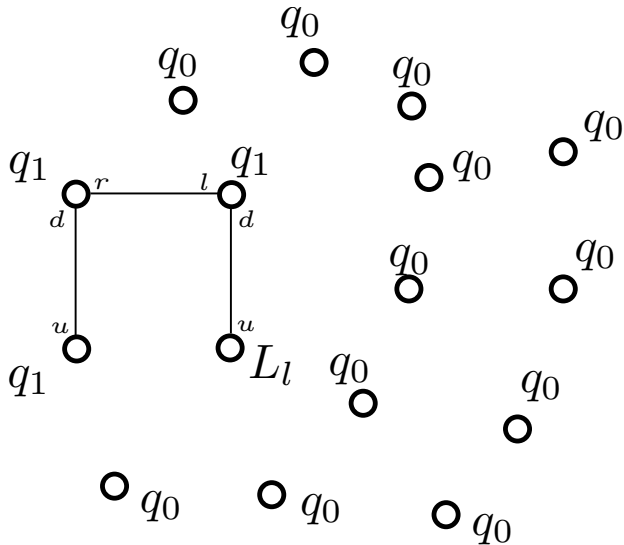


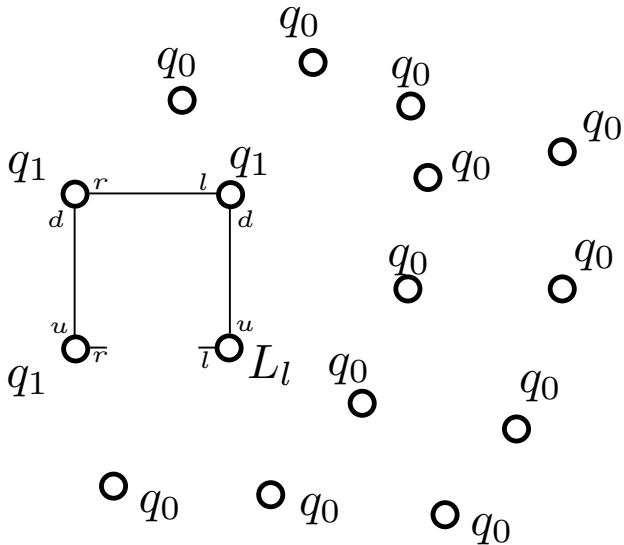


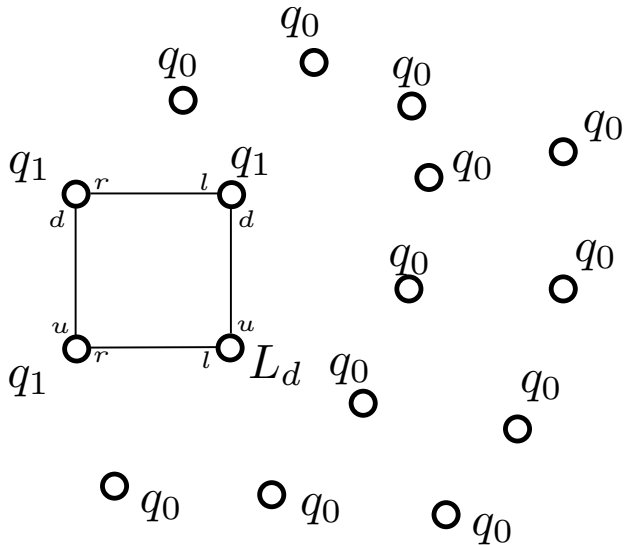


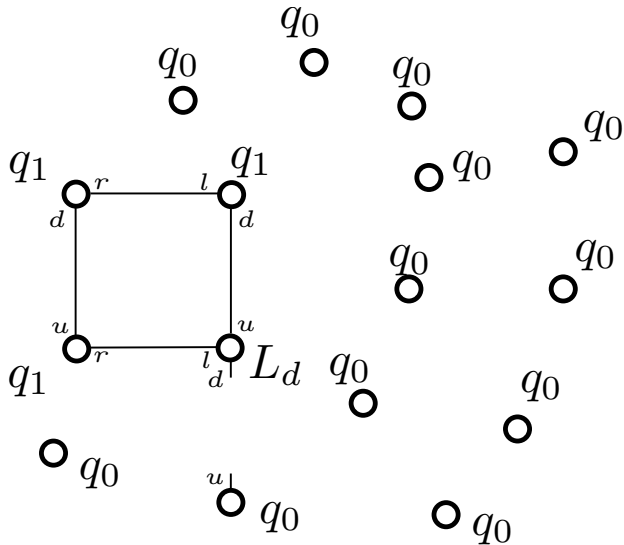


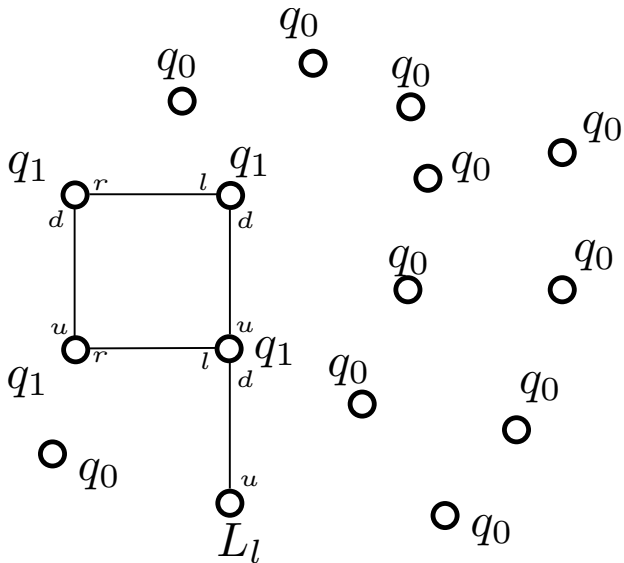


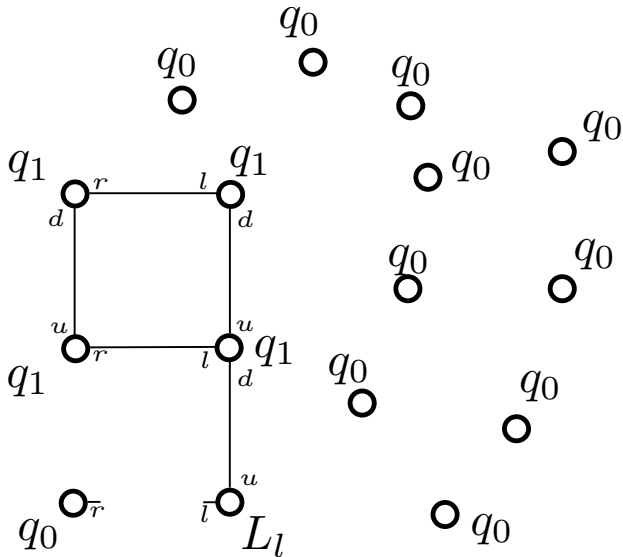


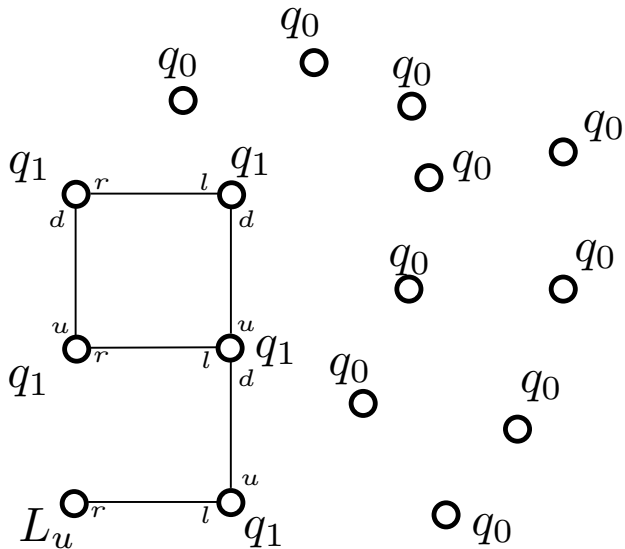


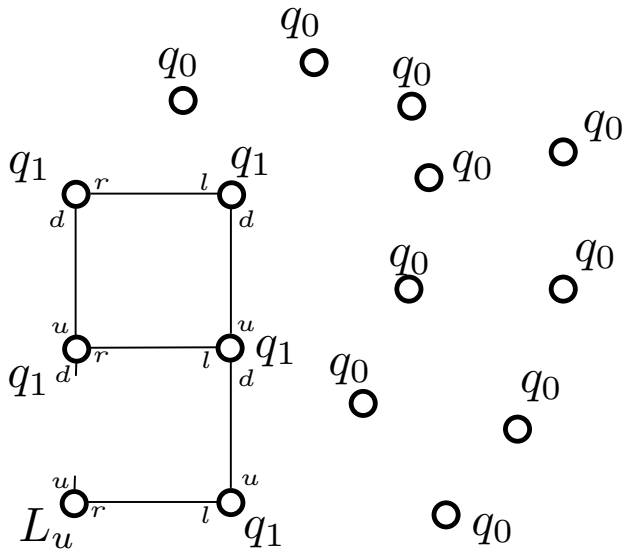


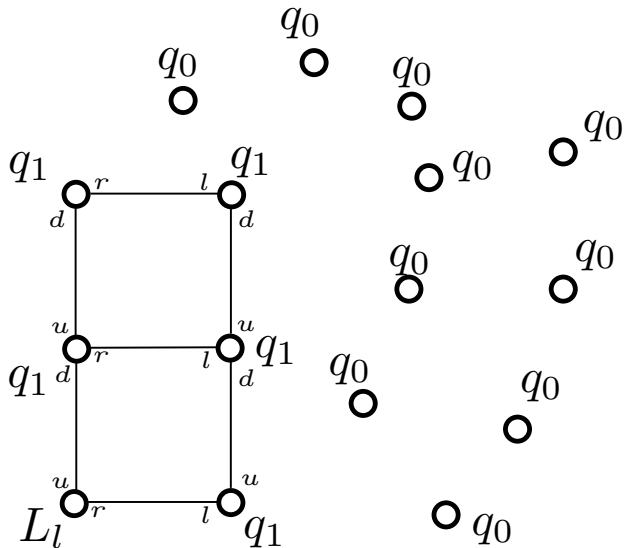


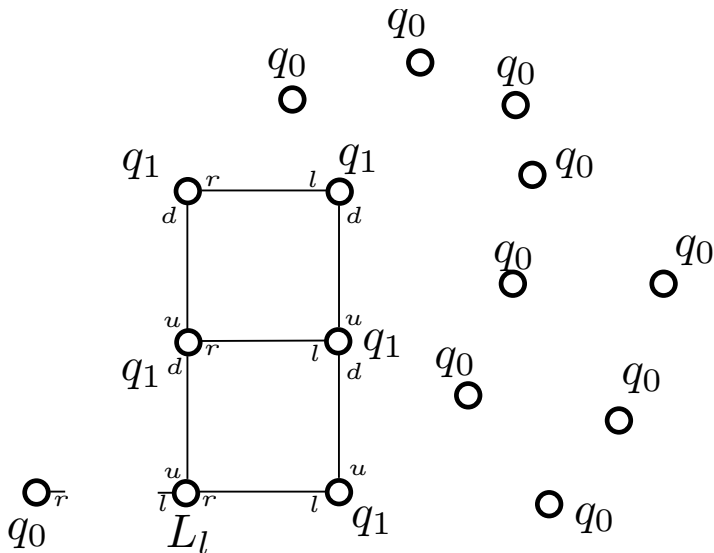


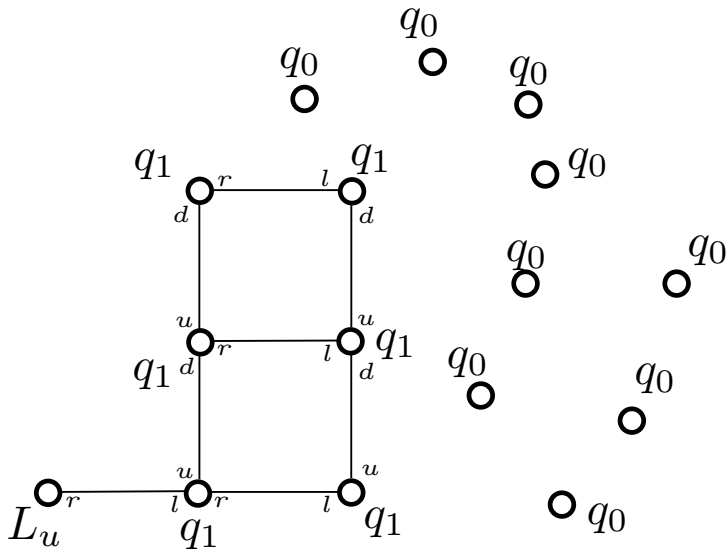


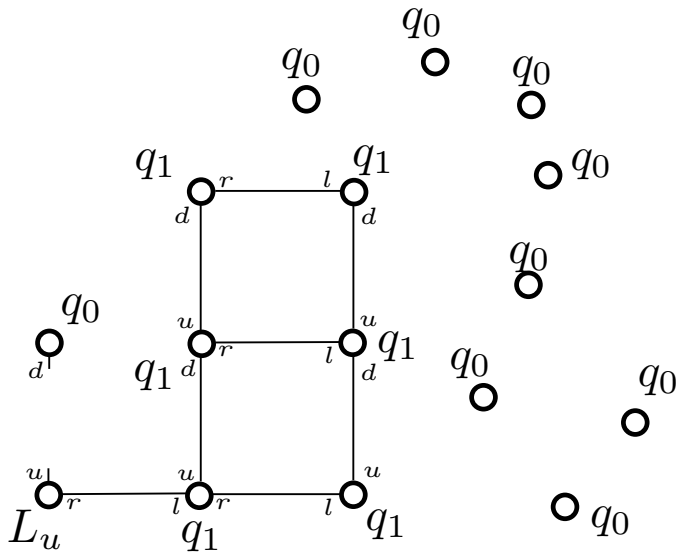


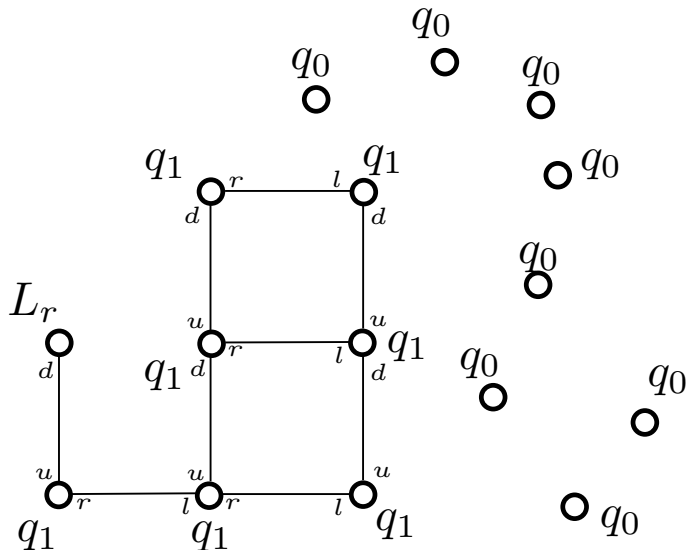


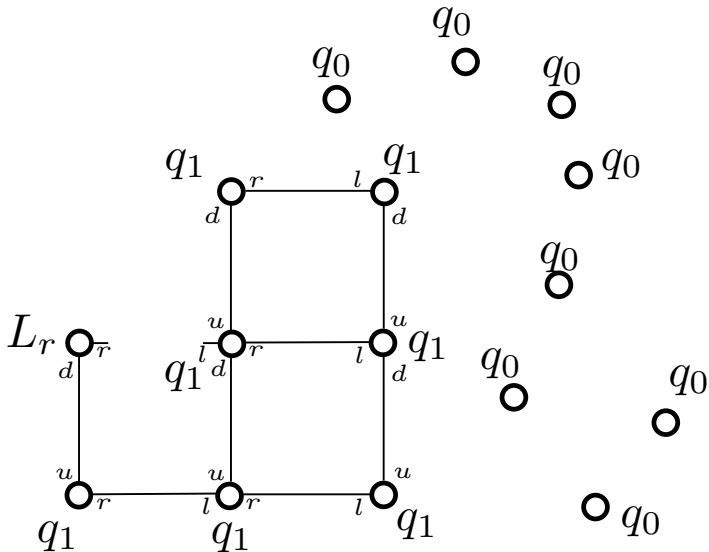


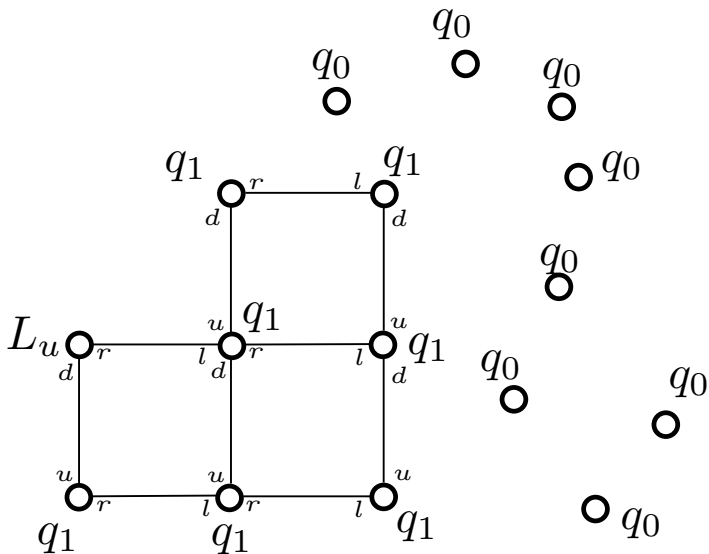


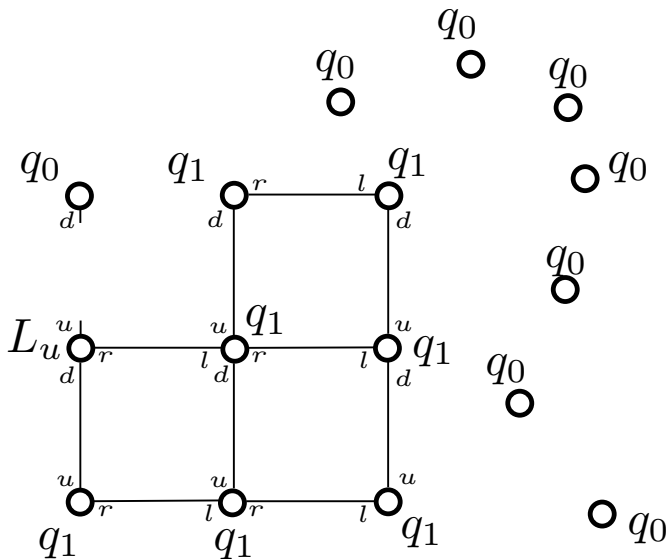


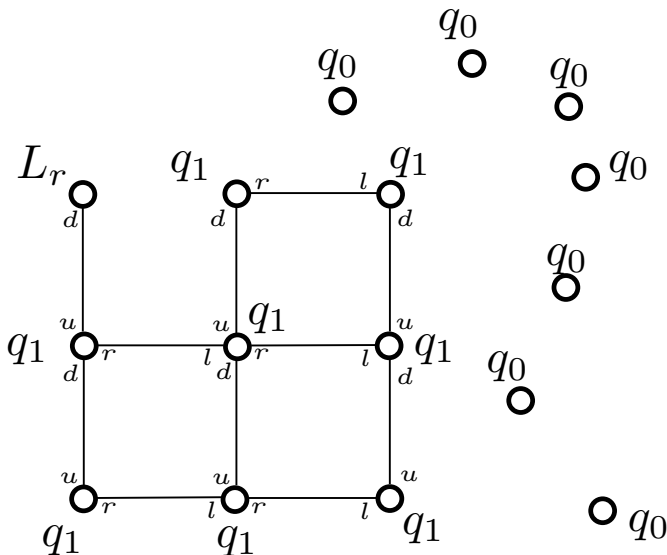


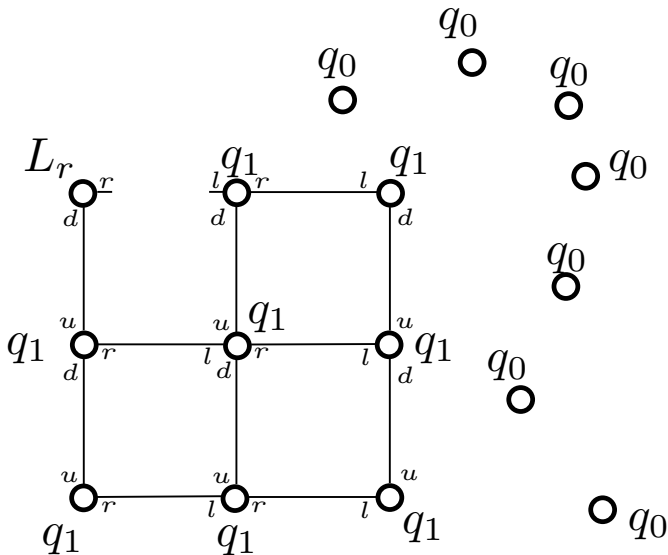


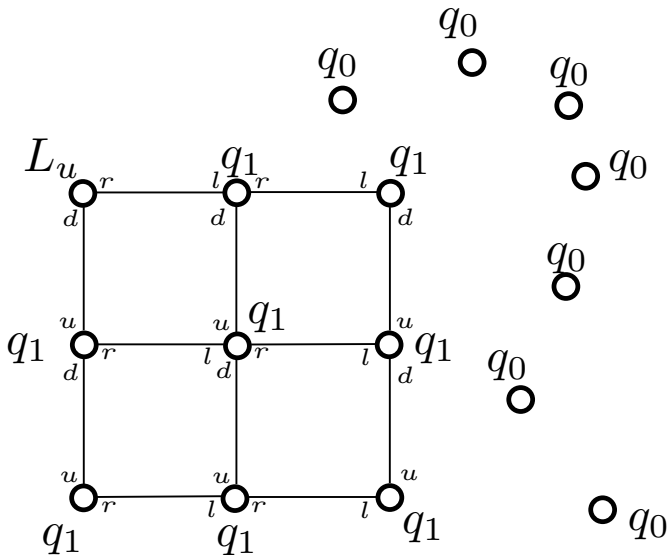


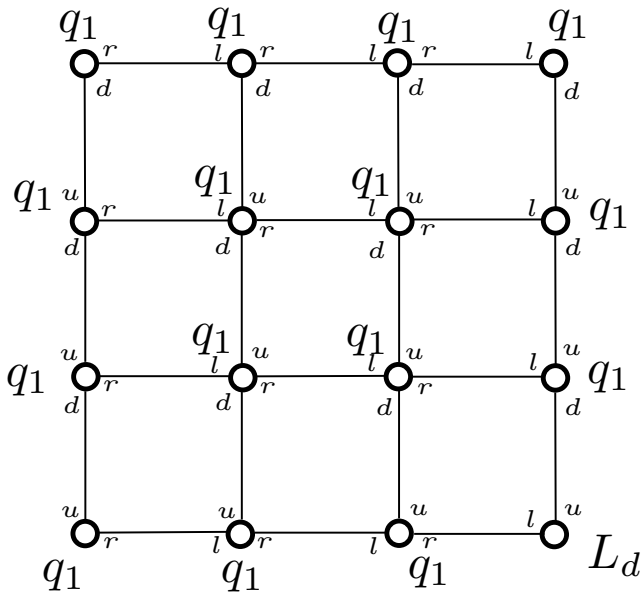












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$$(l(r_0, r_1), q_1) \rightarrow (l(r_0, r_1 + 1), q_2), \text{ and}$$

$$(l(r_0, r_1), \cdot) \rightarrow (\text{halt}, \cdot) \text{ if } r_0 = r_1$$

- r_0 counts the number of q_0 s in the population
- r_1 counts the number of q_1 s in the population
- When a q_0 (q_1) is counted it is converted to q_1 (q_2)
- Terminates when $r_0 = r_1$ for the first time

- $l(r_0, r_1)$: The state of l , where r_0, r_1 are the values of the two counters, $0 \leq r_0, r_1 \leq n$

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Theorem

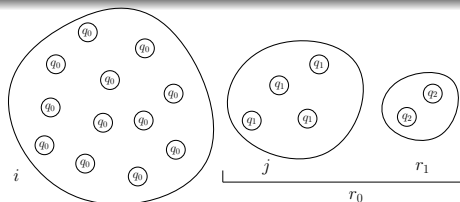
Counting-Upper-Bound halts in every execution. Moreover, if the scheduler is a uniform random one, when this occurs, w.h.p. it holds that $r_0 \geq n/2$.

Proof

- $p_{ij} = i/(i+j)$: probability that an effective interaction is an (l, q_0)
- $q_{ij} = 1 - p_{ij} = j/(i+j)$: probability that it is an (l, q_1)
- r.w. on a line with $n+1$ positions $0, 1, \dots, n$
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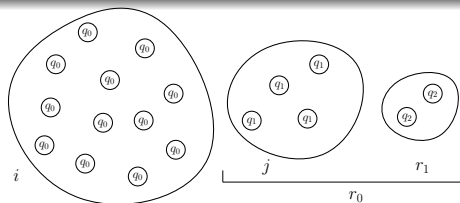


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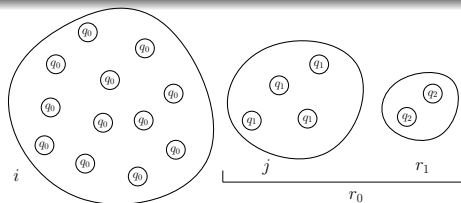


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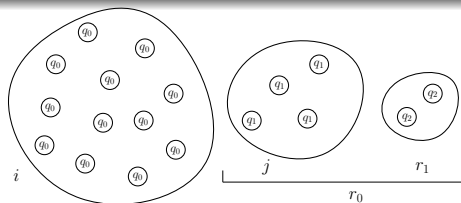


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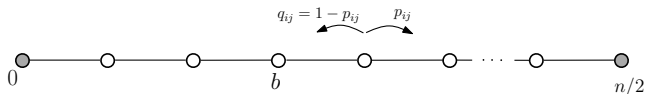
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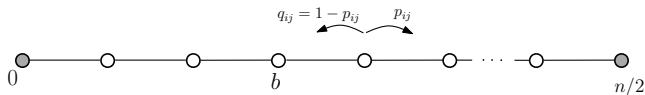
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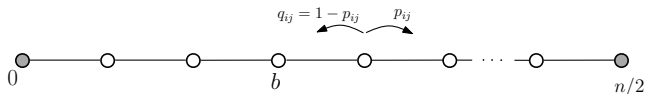
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- “difficult” r.w.: the transition probabilities *depend on the position j and also on $i + j$ which decreases in time*
- upper bound $P[\text{failure}] = P[\text{reach } 0 \text{ before } r_0 \geq n/2 \text{ holds}] \leq$
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- $r_0 + r_1 \leq n$ but $r_1 \leq r_0 \Rightarrow 2r_1 \leq r_0 + r_1$, thus
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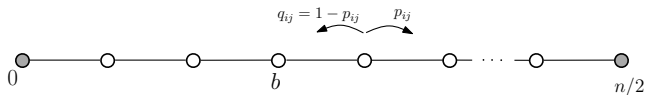
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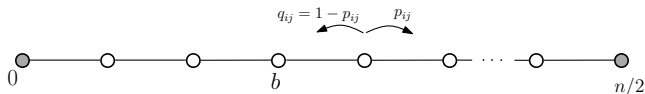
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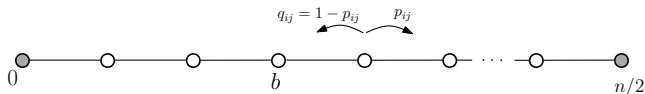
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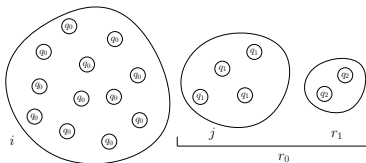
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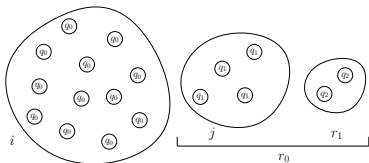
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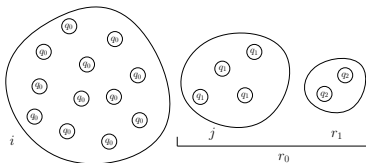
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- If we set $n' = (n/2) - 1$ we have $i + j \geq n'$
- When $r_0 + r_1 = n + 1$ we have $n + 1 = r_0 + r_1 \leq 2r_0 \Rightarrow r_0 \geq n/2$
- During the first n effective interactions: $i + j \geq n' = (n/2) - 1$
- When *interaction $n + 1$* occurs: $r_0 \geq n/2$
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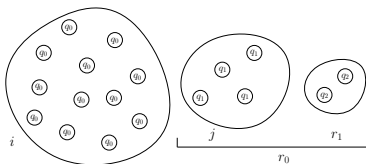
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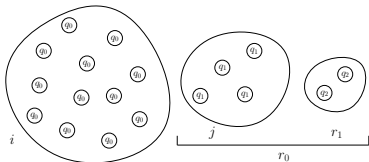
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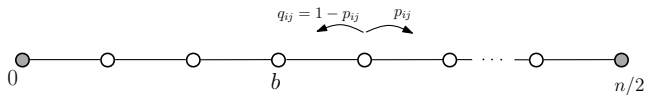
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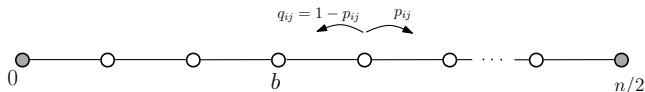
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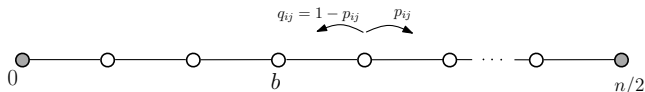
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- $i + j \geq n'$ implies that $p_j \geq (n' - j)/n'$ and $q_j \leq j/n'$
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 - *Gas molecules moving randomly in a container, divided into two urns*
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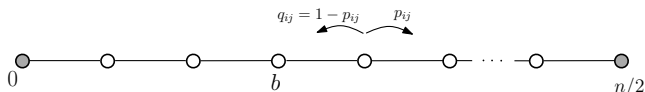
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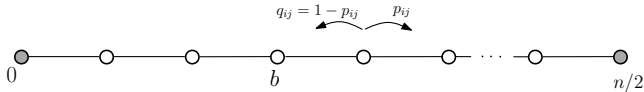
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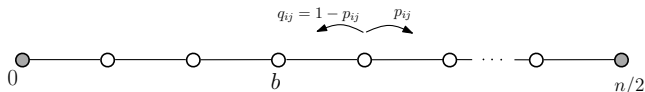
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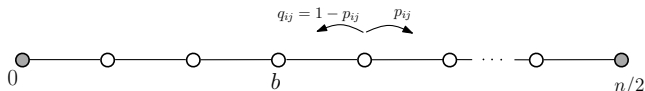
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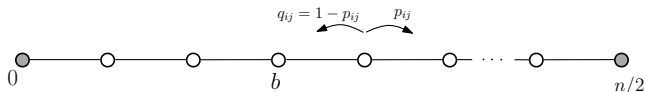
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- *Reduce the Ehrenfest walk to one in which the probabilities **do not depend on j***
- *Further restrict the walk, to the prefix $[0, b]$ of the line*
- *It holds that $j \leq b$, implying that $p \geq (n' - b)/n'$ and $q \leq b/n'$*
- *Set $p = (n' - b)/n'$ and $q = b/n'$*
 - *This may only increase the probability of failure*
- *Imagine now an **absorbing barrier** at 0 and another one at b*
- *Whenever the r.w. is on $b - 1$ it will either return to b before reaching 0 or it will reach 0 (and fail) before returning to b*
- *A r.w. with $b + 1$ positions, where 0 and b are absorbing*
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- **Characterization** for the class of **constructible 2D shape languages**
- Simulate shape-constructing TMs to realize their output-shape in the distributed system
- ① **Counting Protocol**: constructs w.h.p. a line of length $\Theta(\log n)$, containing n in binary
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Theorem

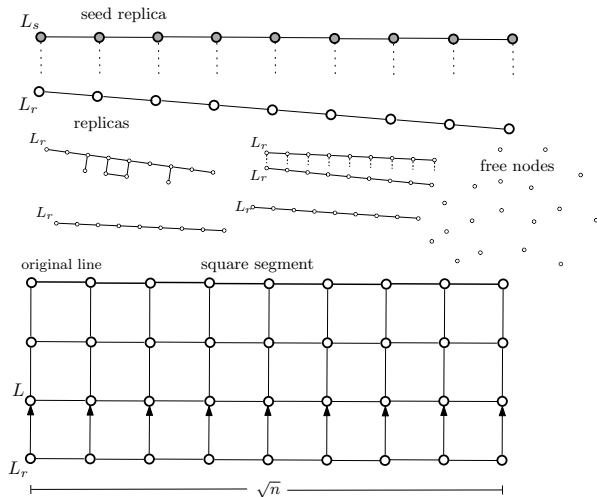
Let $\mathcal{L} = (S_1, S_2, \dots)$ be a connected *2D shape language*, such that \mathcal{L} is *TM-computable in space d^2* . Then there is a protocol that *w.h.p. constructs \mathcal{L}* . In particular, for all $d \geq 1$, whenever the protocol is executed on a population of size $n = d^2$, *w.h.p. it constructs S_d and terminates*. In the worst case, when G_d (that is, the shape of S_d) is a line of length d , the waste is $(d - 1)d = O(d^2) = O(n)$.

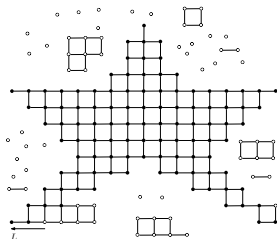
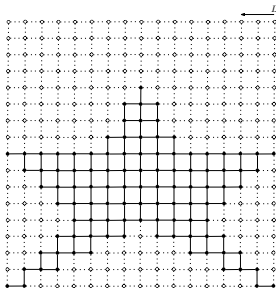
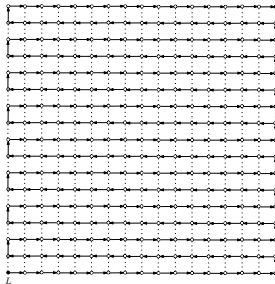
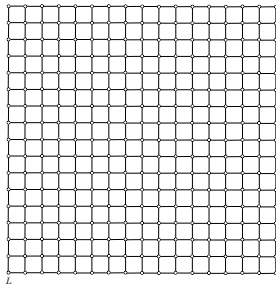
- Adapt Counting-Upper-Bound to work in the present model
- The same probabilistic process
- The leader constructs a line that stores the two counters in binary
 - The line grows whenever more space is required

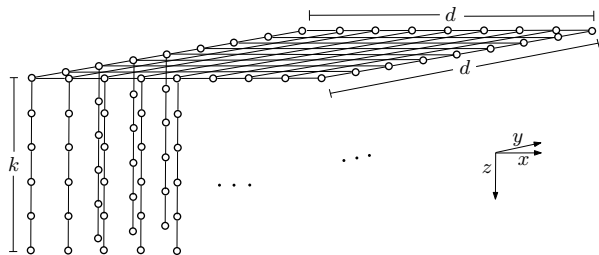
Lemma

Counting-on-a-Line protocol terminates in every execution. Moreover, when the leader terminates, w.h.p. it has formed an active line of length $\log n$ containing n in binary in the r_0 components of the nodes of the line (each node storing one bit).

Constructing a $\sqrt{n} \times \sqrt{n}$ Square







Theorem

Let $\mathcal{L} = (S_1, S_2, \dots)$ be a *TM-computable connected 2D shape language*, such that S_d is computable in space $k = f(d)$ and k is computable in space $O(k \cdot d^2)$. Then there is a protocol that *w.h.p. constructs \mathcal{L}* . In particular, for all $d \geq 1$, whenever the protocol is executed on a population of size $n = k \cdot d^2$, *w.h.p. it constructs S_d and terminates, by executing d^2 simulations in parallel each with space $O(k)$* .

- Complete characterization of constructed networks
- Give a faster than $O(n^3)$ protocol for global line (e.g. $O(n^2 \log n)$)
- Count w.h.p. and terminate if all nodes are initially identical?
- Models of active mobility/actuation (or hybrid active-passive)
- Take other physical considerations into account
 - mass, strength of bonds, rigid and elastic structure, collisions
- Structures that optimize some global property or that achieve a behavior/functionality
- Protocols that efficiently reconstruct broken parts of the structure
- Draw connections to natural processes and to self-assembly and programmable matter models
- We need more real systems-collectives of large numbers of simple interacting devices

Thank You!