

Constant Space and Non-Constant Time in Distributed Computing

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OPODIS
20th December 2017
Lisbon, Portugal

Time complexity versus space complexity

- A well-established topic in centralised complexity theory.
- For example, $\text{NP} \subseteq \text{PSPACE} \subseteq \text{EXP}$.
- What can be said in the distributed setting?
- A message-passing model:
 - problem instance = communication graph + local inputs,
 - time = number of synchronous communication rounds,
 - space = number of bits per node needed to represent the states.

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- Constant time complexity \Rightarrow constant space complexity.
- Does the converse hold?

Time vs. space in the distributed setting

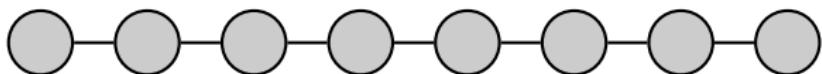
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- More specifically: does there exist a distributed graph problem that is
 - solvable in constant space,
 - not solvable in constant time?

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- Our result: YES, constant space and constant time can be separated!

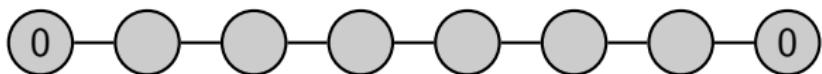
What are the right assumptions? (1/2)

- Easy to construct constant-space non-constant-time problems if
 - promise that the graph is a path, or
 - nodes do not need to halt.
- Count the distance modulo 2 to the nearest degree-1 node:



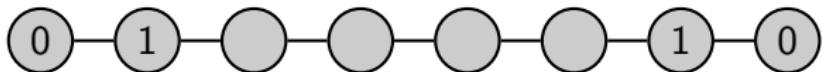
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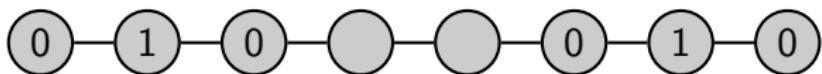
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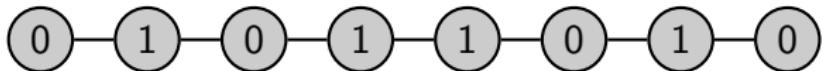
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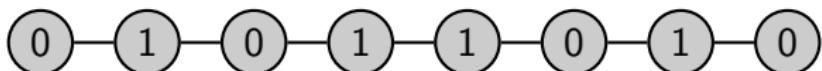
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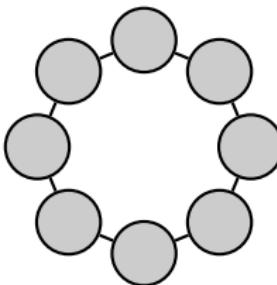


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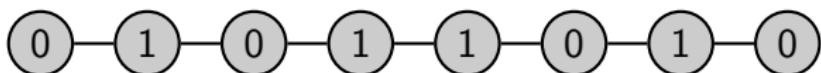


- But what if the input is a cycle?

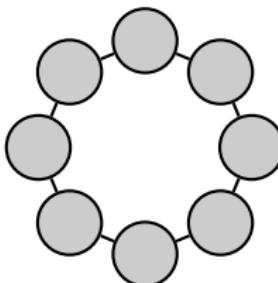


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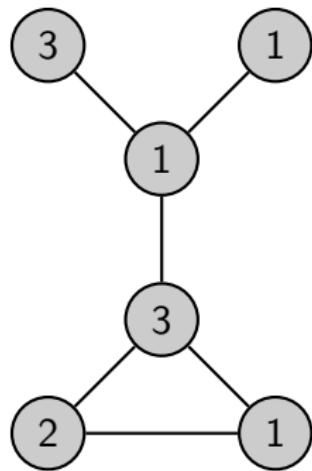


- Our result does not require any promises about the input.

What are the right assumptions? (2/2)

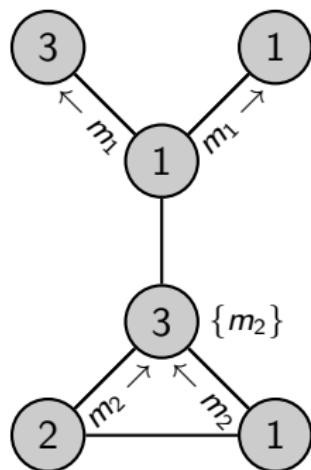
- To achieve a strong separation result, we want a graph problem Π that
 - is solvable in constant space in a very weak model of computation,
 - cannot be solved in constant time even in a very strong model.
- Hence, we will present an algorithm for Π in a very weak model of computation:
 - no unique IDs,
 - no randomness,
 - only constant-size local inputs,
 - only weak communication capabilities.

Model of computation



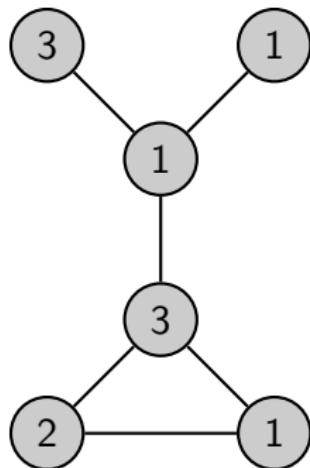
- A simple finite connected undirected graph, with constant-size local inputs.
- An identical deterministic state machine on each node.

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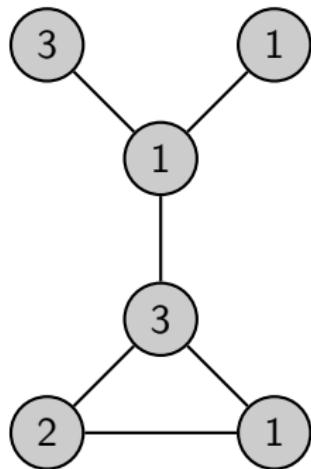
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- Computation proceeds in synchronous rounds:
 - ➊ broadcast a message to neighbours,
 - ➋ receive a set of messages,
 - ➌ set a new state based on previous state and received messages.

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- Computation proceeds in synchronous rounds:
 - ➊ broadcast a message to neighbours,
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 - ➌ set a new state based on previous state and received messages.
- In all graphs, each node eventually halts and produces an output.

Complexity measures



- Given an algorithm (a state machine), its
- running time or time complexity is the number of communication rounds until all nodes have halted,
 - space complexity is the number of bits needed to encode all the states that are visited at least once,
- as a function of n , over all graphs of n nodes.

Our main result

Problem

Construct a graph problem Π such that

- ① *there exists a constant-space algorithm \mathcal{A} that halts and solves Π in all (finite, simple, and connected) graphs, and*
- ② *Π is not solvable by any constant-time algorithm.*

Theorem

There does exist a decision graph problem Π that satisfies the above requirements (1) and (2).

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Theorem (Stronger result)

There does exist a decision graph problem Π that satisfies the above requirements (1) and (2), and that is not solvable by any sublinear-time algorithm even in the class of graphs of maximum degree 2.

An intriguing binary sequence

- The *Thue–Morse sequence* is the infinite sequence (over $\{0, 1\}$) whose prefixes T_i of length 2^i are defined as follows:
 - start with $T_0 = 0$,
 - obtain T_i from T_{i-1} by mapping $0 \mapsto 01$ and $1 \mapsto 10$.
- First steps:

$$T_0 = 0$$

$$T_1 = 01$$

$$T_2 = 0110$$

$$T_3 = 01101001$$

$$T_4 = 0110100110010110$$

⋮

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$$\vdots$$

- Interesting properties:
 - For each $i \in \mathbb{N}$, T_{2i} is a palindrome.
 - The sequence does not contain any cubes, i.e. subwords XXX for any $X \in \{0, 1\}^*$.

Towards a decision graph problem

- Could we separate paths labelled with a prefix T_i from all other paths and cycles by a distributed algorithm?
- The recursive definition of Thue–Morse can be applied backwards
⇒ Given sequence T_i , get back to $T_0 = 0$.
- ... $T_i T_i T_i \dots$ does not appear in the Thue–Morse sequence
⇒ A cycle graph looks different from a path graph.
- A promising idea:
 - Yes-instance: a path labelled with a prefix of the Thue–Morse sequence.
 - No-instance: anything else.

Formalising the idea: the graph problem Π (1/2)

- Define the set of *valid* words over $\{0, 1, _\}$:
 - $_0_\!$ is valid,
 - if X is valid and Y is obtained from X by mapping $0 \mapsto 0_1_1_0$ and $1 \mapsto 1_0_0_1$, then Y is valid.
- The valid words are prefixes of length 4^k of the Thue–Morse sequence, with a separator $_$ added at the beginning, between each pair of consecutive symbols, and at the end.

The decision graph problem Π (2/2)

- Local inputs from $\{A, B, C\} \times \{0, 1, _\}$.
- Local outputs from $\{\text{yes}, \text{no}\}$.
- An instance is a yes-instance if and only if
 - the graph is a path graph,
 - the first parts of the local inputs define a consistent orientation for the path: $\dots ABCABCABC\dots$,
 - the second parts of the local inputs define a *valid* word over $\{0, 1, _\}$.

The algorithm: a high-level idea (1/2)

In each node v of G :

- ① Verify degree and orientation: if $\deg(v) \in \{1, 2\}$ and the orientation is locally consistent, continue; otherwise, reject.
⇒ G is essentially an oriented path, with a port-numbering.

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- ② Verify the input word locally: if every other label is from $\{0, 1\}$ and every other label is $_$, continue; otherwise, reject.
⇒ Copy the input label as the *current label* of v .
⇒ Maintain an invariant: always a separator $_$ at some finite distance.

The algorithm: a high-level idea (2/2)

In each node v of G :

- ③ Apply the recursive definition of Thue–Morse backwards:

$\underbrace{0+1+1+0+1+0+0+1+}_{\Downarrow} \quad \underbrace{1+0+0+1+0+1+1+0+}_{\Downarrow}$
 0000000000+ 1111111111+ 1111111111+ 0000000000+

If the pattern does not match or the new label for v is ambiguous, reject; otherwise, repeat.

⇒ The invariant is maintained.

⇒ The word encoded in the path goes consistently from T_{2j} to $T_{2(j-1)}$.

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- ④ If the word matches $|_0+_1|$ or $|_0+_1+_1+_0+_|$, accept.
(Here $|$ denotes the end of the path.)

Examples (1/3)

- Path graph, yes-instance:

_0_1_1_0_1_0_0_1_1_0_0_1_0_1_1_0_
 ↓ (unambiguous substitutions)
_0000000_1111111_1111111_0000000_
 ↓
accept

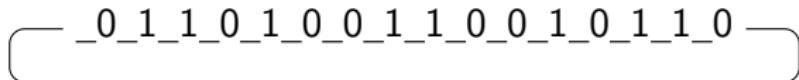
Examples (2/3)

- Path graph, no-instance:

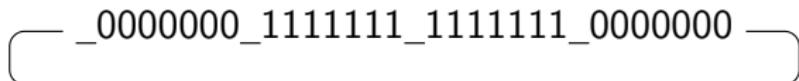
_0_1_1_0_1_0_0_1_1_0_1_0_0_1_
 ↓
_0000000_1111111_...
 ..._0000000_1111111_
 ↓ (ambiguous substitutions)
 reject

Examples (3/3)

- Cycle graph:

 _0_1_1_0_1_0_0_1_1_0_0_1_0_1_1_0

⇓ (unambiguous substitutions)

 _0000000_1111111_1111111_0000000

⇓ (no matches)

reject

Complexity

- The substitutions involve constant number of blocks separated by _'s
⇒ constant space is enough.
- Need to receive information from the other end of the path
⇒ $\Omega(n)$ time is needed – even if we have unique IDs or randomness.
- Substitution phase i takes $O(c^i)$ rounds (c constant), $O(\log n)$ phases
⇒ $O(n)$ time is enough.

Conclusion

- We proved a strong separation between constant space and constant time by introducing a graph problem that
 - can be solved in constant space in a very limited model,
 - requires linear time in strong models (e.g. LOCAL with randomness).
- However, our problem is highly artificial. It is open, whether there exist
 - natural graph problems, or
 - LCL (locally checkable labelling) problemswith the above properties.

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Thanks! Questions?