A tight lower bound for simulating multiset model by set model in distributed computing

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Distributed system



- A graph, whose each node
 - runs the same algorithm,
 - can communicate with its neighbours,
 - produces a local output.



In every round, each node v

sends messages to its neighbours,

- receives messages from its neighbours,
- updates its state.



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After the final round, each node announces its own output.

Focus on communication, not computation



The running time of an algorithm is the *number of communications rounds*.

The running time may depend on

- the maximum degree of the graph, Δ ,
- the number of nodes, n.

Port numbering

A port of a graph G = (V, E) is a pair (v, i), where $v \in V$ and $i \in \{1, 2, ..., \deg(v)\}$. Let P(G) be the set of all ports of G. A port numbering of G is a bijection $p: P(G) \rightarrow P(G)$ such that

p(v,i) = (u,j) for some *i* and *j* if and only if $\{v, u\} \in E$.

Intuitively, if p(v, i) = (u, j), then (v, i) is an output port of node v that is connected to an input port (u, j) of node u.





We say that a port numbering p is *consistent* if we have

$$p(p(v,i)) = (v,i)$$
 for all $(v,i) \in P(G)$,

or, in other words, if the input port and the output port connected to the same neighbour always have the same number.



$$p(u, 1) = (v, 2),$$

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An *input* for a graph G = (V, E) is a function $f: V \to X$, where $X \ni \emptyset$ is a finite set. For each $v \in V$, the value f(v) is called the *local input* of v.

The symbol $\emptyset \in X$ is used to indicate "no input".

Algorithms as state machines

Let $\Delta \in \mathbb{N}_+$ and let X be a set of local inputs. A *distributed state machine* for $(\mathcal{F}(\Delta), X)$ is a tuple $\mathcal{A} = (Y, Z, \sigma_0, M, \mu, \sigma)$, where

- Y is a set of states,
- $Z \subseteq Y$ is a finite set of stopping states,
- $\sigma_0 \colon \{0, 1, \dots, \Delta\} \times X \to Y$ is a function that defines the initial state,
- M is a set of messages such that $\epsilon \in M$,
- μ: Y × [Δ] → M is a function that constructs the outgoing messages, such that μ(z, i) = ε for all z ∈ Z and i ∈ [Δ],
- $\sigma: Y \times M^{\Delta} \to Y$ is a function that defines the state transitions, such that $\sigma(z, \overline{m}) = z$ for all $z \in Z$ and $\overline{m} \in M^{\Delta}$.

The special symbol $\epsilon \in M$ indicates "no message".

Execution

Let $G = (V, E) \in \mathcal{F}(\Delta)$, let p be a port numbering of G, let $f : V \to X$, and let \mathcal{A} be a distributed state machine for $(\mathcal{F}(\Delta), X)$.

The state of the system in round $r \in \mathbb{N}$ is a function $x_r \colon V \to Y$, where $x_r(v)$ is the *state* of node v in round r. To initialise the nodes, set

 $x_0(v) = \sigma_0(\deg(v), f(v))$ for each $v \in V$.

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Then, assume that x_r is defined for some $r \in \mathbb{N}$. Let $(u, j) \in P(G)$ and (v, i) = p(u, j). Now, node v receives the message

$$a_{r+1}(v,i) = \mu(x_r(u),j)$$

from its port (v, i) in round r + 1. For each $v \in V$, we define

$$\overline{a}_{r+1}(v) = (a_{r+1}(v,1), a_{r+1}(v,2), \dots, a_{r+1}(v, \deg(v)), \epsilon, \epsilon, \dots, \epsilon) \in M^{\Delta}$$

Now we can define the new state of each node $v \in V$ as follows:

$$x_{r+1}(v) = \sigma(x_r(v), \overline{a}_{r+1}(v)).$$

Let $t \in \mathbb{N}$. If $x_t(v) \in Z$ for all $v \in V$, we say that \mathcal{A} stops in time t in (G, f, p).

The running time of A in (G, f, p) is the smallest t for which this holds.

If \mathcal{A} stops in time t in (G, f, p), the *output* of \mathcal{A} in (G, f, p) is $x_t \colon V \to Y$.

For each $v \in V$, the *local output* of v is $x_t(v)$.

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Let X and Y be finite nonempty sets.

A graph problem is a function $\Pi_{X,Y}$ that maps each undirected simple graph G = (V, E) and each input $f \colon V \to X$ to a set $\Pi_{X,Y}(G, f)$ of solutions.

Each solution $S \in \Pi_{X,Y}(G, f)$ is a function $S \colon V \to Y$.

Solving a graph problem

Let $\Pi_{X,Y}$ be a graph problem, $T : \mathbb{N} \times \mathbb{N} \to \mathbb{N}$ and $\mathbf{A} = (\mathcal{A}_1, \mathcal{A}_2, ...)$ such that each \mathcal{A}_Δ is a distributed state machine for $(\mathcal{F}(\Delta), X)$. Algorithm \mathbf{A} solves $\Pi_{X,Y}$ in time T if the following holds for all $\Delta \in \mathbb{N}$, all finite graphs $G = (V, E) \in \mathcal{F}(\Delta)$, all inputs $f : V \to X$ and all port numberings p of G:

- \mathcal{A}_{Δ} stops in time $T(\Delta, |V|)$ in (G, f, p).
- **2** The output of \mathcal{A}_{Δ} in (G, f, p) is in $\Pi_{X,Y}(G, f)$.

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- **2** The output of \mathcal{A}_{Δ} in (G, f, p) is in $\Pi_{X,Y}(G, f)$.

We say that **A** solves $\Pi_{X,Y}$ in time *T* assuming consistency if the above holds for all consistent port numberings *p* of *G*.

If $T(\Delta, n)$ does not depend on n, we say that **A** solves $\Pi_{X,Y}$ in constant time or that **A** is a local algorithm for $\Pi_{X,Y}$.



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Example problems:

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Example problems:

- minimum vertex cover,
- maximal matching.

 $\mathcal{V}\mathcal{V}$ is the class of all distributed state machines (send a vector, receive a vector).

We can place different restrictions on the algorithms:

 $\bullet \ \mathcal{VB}:$ broadcast the same message to all neighbours:

$$\mu(y,i)=\mu(y,j)$$
 for all $i,j\in\{1,2,\ldots,\Delta\}$ and $y\in Y,$

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• \mathcal{MV} : receive a multiset of messages:

 $\mathsf{multiset}(\overline{a}) = \mathsf{multiset}(\overline{b}) \Rightarrow \sigma(y, \overline{a}) = \sigma(y, \overline{b}) \text{ for all } y \in Y,$

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•
$$\mathcal{MB} = \mathcal{MV} \cap \mathcal{VB}$$
 and $\mathcal{SB} = \mathcal{SV} \cap \mathcal{VB}$.

$$\begin{split} \mathbf{VV} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{VV} \text{ for all } \Delta \}, \\ \mathbf{MV} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{MV} \text{ for all } \Delta \}, \\ \mathbf{SV} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{SV} \text{ for all } \Delta \}, \\ \mathbf{VB} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{VB} \text{ for all } \Delta \}, \\ \mathbf{MB} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{MB} \text{ for all } \Delta \}, \\ \mathbf{SB} &= \{ (\mathcal{A}_1, \mathcal{A}_2, \dots) : \mathcal{A}_\Delta \in \mathcal{SB} \text{ for all } \Delta \}. \end{split}$$

Let P be the class of all graph problems.

$$\begin{split} VV_c &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{VV} \ \text{that solves } \Pi \ \text{assuming consistency} \}, \\ VV &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{VV} \ \text{that solves } \Pi \}, \\ MV &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{MV} \ \text{that solves } \Pi \}, \\ SV &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{SV} \ \text{that solves } \Pi \}, \\ VB &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{VB} \ \text{that solves } \Pi \}, \\ MB &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{MB} \ \text{that solves } \Pi \}, \\ SB &= \{\Pi \in P \ : \ \text{there is } \textbf{A} \in \textbf{SB} \ \text{that solves } \Pi \}. \end{split}$$

Containment relations between the classes



Trivial relations:

- $\bullet \ \mathsf{SV} \subseteq \mathsf{MV} \subseteq \mathsf{VV} \subseteq \mathsf{VV}_{\mathsf{c}}\text{,}$
- $\bullet \ \mathsf{SB} \subseteq \mathsf{MB} \subseteq \mathsf{VB},$
- $VB \subseteq VV$,
- $\bullet \ \mathsf{MB} \subseteq \mathsf{MV},$
- $SB \subseteq SV$.

Non-trivial: $SV \subseteq VB$? $VB \subseteq SV$?

Containment relations between the classes



Hella, Järvisalo, Kuusisto, Laurinharju, L., Luosto, Suomela, Virtema (PODC 2012):

$$SB \subsetneq MB = VB \subsetneq SV = MV = VV \subsetneq VV_c$$
.

Simulation results



Equalities are proved by showing that seemingly more powerful algorithms can be simulated by seemingly weaker algorithms.

- MV = VV and MB = VB: the simulation does not increase running time.
- SV = MV: the running time increases by $2\Delta 2$ rounds.

New contributions



SV = MV: the running time increases by $2\Delta - 2$ rounds.

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Question: Is the overhead of $2\Delta - 2$ rounds optimal?

Answer: Yes

- A so-called simulation problem requires exactly $2\Delta 2$ rounds.
- In the case of graph problems, a linear-in-Δ overhead is necessary.

Trivially $SV \subseteq MV$.

Hella, Järvisalo, Kuusisto, Laurinharju, L., Luosto, Suomela, Virtema (PODC 2012):

Theorem

Let Π be a graph problem and let $T : \mathbb{N} \times \mathbb{N} \to \mathbb{N}$. Assume that there is an algorithm $\mathbf{A} \in \mathbf{MV}$ that solves Π in time T. Then there is an algorithm $\mathbf{B} \in \mathbf{SV}$ that solves Π in time T', where $T'(n, \Delta) = T(n, \Delta) + 2\Delta - 2$.

It follows that SV = MV.

Idea behind the simulation theorem

First, solve the following simulation problem by an \mathcal{SV} -algorithm:



If $p_1 = p_2$, then output(v) \neq output(w).

Now the pair

(output, port number)

is distinct for each neighbour.

This takes $2\Delta - 2$ communication rounds.

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This takes $2\Delta - 2$ communication rounds.

Then, simulate the \mathcal{MV} -algorithm by attaching the above pair to each message. That way we can reconstruct the message multiplicities.

The new results: lower bounds for the simulation

Is the overhead of $2\Delta-2$ rounds really needed to reconstruct the message multiplicities by an $\mathcal{SV}\text{-algorithm}?$

Theorem

For each $\Delta \ge 2$ there is a graph $G = (V, E) \in \mathcal{F}(\Delta)$, a port numbering p of G and nodes v, u, $w \in V$ such that when executing any algorithm $\mathcal{A} \in S\mathcal{V}$ in (G, p), node v receives identical messages from its neighbours u and w in rounds $1, 2, \ldots, 2\Delta - 2$.
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Theorem

There is a graph problem Π that can be solved in one round by an algorithm in **MV** but that requires at least time T, where $T(n, \Delta) \ge \Delta$ for all $\Delta \ge 2$, when solved by an algorithm in **SV**.

Example: a problem instance separating \mathcal{SV} and \mathcal{MV}



Output 1 if there is an even number of neighbours of even degree, 0 otherwise.

Lower-bound construction for the simulation problem



- Investigate walks that start from the blue nodes and follow an identical sequence of port numbers.
 - In which cases we cannot extend the walks in a consistent manner?
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 - In which cases we cannot extend the walks in a consistent manner?
 - What is the length of such maximal walks?
- Prove a lower bound for the length of the walks.
- Show that the lower bound on walks implies bisimilarity of the blue nodes up to a certain distance.
- Bisimilarity entails a lower bound for the running time of any distributed algorithm that is able to distinguish the nodes.

If p(v, i) = (u, j), we write $\pi(v, u) = i$. That is, $\pi(v, u)$ is the number of the output port of v that is connected to u.

Definition of the graph G_d

$$egin{aligned} c_1^j &= \min(\{1,2,\ldots,d\} \setminus \{b_2^+,c_1^1,c_1^2,\ldots,c_1^{j-1}\}), \ c_2^j &= \min(\{1,2,\ldots,d\} \setminus \{b_1,c_2^1,c_2^2,\ldots,c_2^{j-1}\}). \end{aligned}$$

• If $(a_1, a_2, \ldots, a_i) \in V_d$, where *i* is even and 0 < i < 2d, then $(a_1, a_2, \ldots, a_{i+1}^j) \in V_d$ for all $j = 1, 2, \ldots, d-1$, where $a_{i+1}^j = (c_1^j, c_2^j)$ is defined as follows. Let $(b_1, b_2) = a_i$. Define

$$egin{aligned} c_1^j &= \min(\{1,2,\ldots,d\} \setminus \{b_2,c_1^1,c_1^2,\ldots,c_1^{j-1}\}), \ c_2^j &= \min(\{0,1,\ldots,d-1\} \setminus \{b_1,c_2^1,c_2^2,\ldots,c_2^{j-1}\}). \end{aligned}$$

Graph G_d for d = 4



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The set E_d of edges consists of all pairs $\{v, u\}$, where $v = (a_1, a_2, \ldots, a_i) \in V_d$ and $u = (a_1, a_2, \ldots, a_i, a_{i+1}) \in V_d$ for some $i \in \{0, 1, \ldots\}$.

If $v = (a_1, a_2, ..., a_i)$ and $u = (a_1, a_2, ..., a_{i+1})$, where $a_{i+1} = (b_1, b_2)$, the outgoing port number from v to u is $\pi_d(v, u) = b_1$ and the outgoing port number from u to v is $\pi_d(u, v) = b_2$.

Pairs of separating walks (PSWs)

A walk is a sequence $\overline{v} = (v_0, v_1, \dots, v_k)$ of nodes such that $\{v_i, v_{i+1}\} \in E_d$ for all $i = 0, 1, \dots, k - 1$.

A pair $(\overline{v}_1, \overline{v}_2)$ of walks, where $\overline{v}_i = (v_0^i, v_1^i, \dots, v_k^i)$ for all i = 1, 2, is called a *pair of separating walks (PSW) of length k in G_d* if the following conditions hold:

•
$$v_0^1 = ((1,0)) \text{ and } v_0^2 = ((2,1)).$$

• $\pi_d(v_j^1, v_{j-1}^1) = \pi_d(v_j^2, v_{j-1}^2) \text{ for all } j = 1, 2, \ldots, k.$

• There is $v_{k+1}^1 \in V_d$ with $\{v_k^1, v_{k+1}^1\} \in E_d$ such that there is no $v_{k+1}^2 \in V_d$ for which $\{v_k^2, v_{k+1}^2\} \in E_d$ and $\pi_d(v_{k+1}^1, v_k^1) = \pi_d(v_{k+1}^2, v_k^2).$

We say that a pair of separating walks of length k in G_d is critical if there does not exist a pair of separating walks of length k' in G_d for any k' < k.















Definition

Let G = (V, E) and G' = (V', E') be graphs, let f and f' be inputs for Gand G', respectively, and let p and p' be port numberings of G and G', respectively. An r-SV-bisimulation between nodes $v \in V$ and $v' \in V'$ is a sequence of binary relations $B_r \subseteq B_{r-1} \subseteq \cdots \subseteq B_0 \subseteq V \times V'$ such that the following conditions hold for $1 \leq i \leq r$:

$$(v, v') \in B_r.$$

3 If $(u, u') \in B_0$, then $\deg_G(u) = \deg_{G'}(u')$ and f(u) = f'(u').

- If $(u, u') \in B_i$ and $\{u, w\} \in E$, then there is $w' \in V'$ such that $\{u', w'\} \in E'$, $(w, w') \in B_{i-1}$ and $\pi(w, u) = \pi'(w', u')$.
- If $(u, u') \in B_i$ and $\{u', w'\} \in E'$, then there is $w \in V$ such that $\{u, w\} \in E$, $(w, w') \in B_{i-1}$ and $\pi(w, u) = \pi'(w', u')$.

We say that $v \in V$ and $v' \in V'$ are r-SV-bisimilar and write $(G, f, v, p) \bigoplus_{r}^{SV} (G', f', v', p')$ (or simply $v \bigoplus_{r}^{SV} v'$) if there exists an r-SV-bisimulation between them.

Lemma

Let G = (V, E) and G' = (V', E') be graphs, let f and f' be inputs for Gand G', respectively, and let p and p' be port numberings of G and G', respectively. If $(G, f, v, p) \bigoplus_{r}^{SV} (G', f', v', p')$ for some $r \in \mathbb{N}$, $v \in V$ and $v' \in V'$, then for all algorithms $A \in SV$ we have $x_t(v) = x'_t(v')$ for all $t = 0, 1, \ldots, r$, that is, the state of v and v' is identical in rounds $0, 1, \ldots, r$.

Bisimilarity

Lemma

The r-SV-bisimilarity relation \bigoplus_{r}^{SV} is an equivalence relation in the class of quadruples (G, f, v, p), where G = (V, E) is a graph, f is an input for G, p is a port numbering of G and $v \in V$.

Lemma

Let G = (V, E) and G' = (V', E') be graphs, let f and f' be inputs for G and G', respectively, let p and p' be port numberings of G and G', respectively, and let $v \in V$, $v' \in V'$. Then $(G, f, v, p) \bigoplus_{r}^{SV} (G', f', v', p')$ iff the following conditions hold:

- $(G, f, v, p) \underset{r-1}{\overset{\mathcal{SV}}{\leftarrow}} (G', f', v', p').$
- ② If {v, w} ∈ E, then there is w' ∈ V' such that {v', w'} ∈ E', (G, f, w, p) $\underset{r-1}{\overset{SV}{\rightharpoonup}}$ (G', f', w', p') and $\pi(w, v) = \pi'(w', v')$.
- If $\{v', w'\} \in E'$, then there is $w \in V$ such that $\{v, w\} \in E$, $(G, f, w, p) \stackrel{SV}{\to}_{r-1}^{SV} (G', f', w', p')$ and $\pi(w, v) = \pi'(w', v')$.

If $v = (a_1, a_2, ..., a_i)$ and $u = (a_1, a_2, ..., a_{i+1})$, we say that node v is the *parent* of node u and that u is a *child* of v.

We say that the node v is *even* if i is even and *odd* if i is odd.

If $a_i = (b_1, b_2)$, we call (b_1, b_2) the *type* of node v.

Lemma

For each d, we have $\deg(v) \in \{1, d\}$ for all $v \in V_d$, and thus $G_d \in \mathcal{F}(d)$. Additionally, G_d is a subgraph of G_{d+1} .

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Lemma

Let $v \in V_d$ and $a \in \{0, 1, ..., d\}$. Then there is at most one node $u \in V_d$ such that $\{v, u\} \in E_d$ and $\pi_d(u, v) = a$.

Easy observations

Lemma

Let $v = (a_1, a_2, ..., a_i) \in V_d$, where i < 2d. If v is odd, then for all $a \in \{1, 2, ..., d\}$ there exists $u \in V_d$ such that $\{v, u\} \in E_d$ and $\pi_d(u, v) = a$. If v is even, then either for all $a \in \{0, 1, ..., d-1\}$ or for all $a \in \{0, 1, ..., d-2, d\}$ there exists $u \in V_d$ such that $\{v, u\} \in E_d$ and $\pi_d(u, v) = a$. In the case of even v and a = d, node u is the parent of node v.

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Lemma

Let $\{v, u\} \in E_{d+1} \setminus E_d$ be such that $v \in V_d$. Then u is a child of v. If v is odd, then $\pi_{d+1}(v, u) = \pi_{d+1}(u, v) = d + 1$. If v is even, then $\pi_{d+1}(v, u) = d + 1$ and $\pi_{d+1}(u, v) \in \{d - 1, d\}$.

Lemma

Let $(\overline{v}_1, \overline{v}_2)$, where $\overline{v}_i = (v_0^i, v_1^i, \dots, v_k^i)$ for some $k \leq 2d - 3$ and all i = 1, 2, be a PSW in G_d . If for some $\ell \in \{0, 1, \dots, k - 1\}$ the node $v_{\ell+1}^i$ is a child of node v_{ℓ}^i for all i = 1, 2, and we have $\pi_d(v_{\ell}^1, v_{\ell+1}^1) = \pi_d(v_{\ell}^2, v_{\ell+1}^2)$, then $(\overline{v}_1, \overline{v}_2)$ is not a critical PSW in G_d .

Lemma

Let $(\overline{v}_1, \overline{v}_2)$ be a PSW of length $k \leq 2d - 3$ in G_d . Then there is a PSW of length k + 2 in G_{d+1} .

Second-to-last node is in $V_d \setminus V_{d-1}$

Lemma

Let $(\overline{v}_1, \overline{v}_2)$, where $\overline{v}_i = (v_0^i, v_1^i, \dots, v_k^i)$ for some $k \leq 2d - 3$ and all i = 1, 2, be a critical PSW in G_d . Then we have $v_{k-1}^i \in V_d \setminus V_{d-1}$ for some $i \in \{1, 2\}$.

Lemma

Let $(\overline{v}_1, \overline{v}_2)$, where $\overline{v}_i = (v_0^i, v_1^i, \dots, v_k^i)$ for some $k \leq 2d - 3$ and all i = 1, 2, be a pair of walks in G_d such that conditions (1) and (2) hold. If $(\overline{v}_1, \overline{v}_2)$ is not a PSW in G_d , then for each neighbour $v_{k+1}^1 \in V_d$ of v_k^1 there is a neighbour $v_{k+1}^2 \in V_d$ of v_k^2 such that $\pi_d(v_{k+1}^1, v_k^1) = \pi_d(v_{k+1}^2, v_k^2)$, and vice versa.

Lemma

Let $(\overline{v}_1, \overline{v}_2)$, where $\overline{v}_i = (v_0^i, v_1^i, \dots, v_k^i)$ for some $k \le 2d - 3$ and all i = 1, 2, be a critical PSW in G_d . Then $(\overline{v}'_1, \overline{v}'_2)$, where $\overline{v}'_i = (v_0^i, v_1^i, \dots, v_{k-2}^i)$ for all i = 1, 2, is a PSW in G_{d-1} .

Minimum length of a PSW and bisimilarity

Lemma

Let $(\overline{v}_1, \overline{v}_2)$ be a PSW of length $k \leq 2d - 3$ in G_d . Then $k \geq 2d - 3$.

Minimum length of a PSW and bisimilarity

Lemma

Let $(\overline{v}_1, \overline{v}_2)$ be a PSW of length $k \leq 2d - 3$ in G_d . Then $k \geq 2d - 3$.

Lemma

We have $((1,0)) \leftrightarrow_{2d-3}^{SV} ((2,1))$, that is, the nodes ((1,0)) and ((2,1)) of G_d are (2d-3)-SV-bisimilar.

Separation by a graph problem



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Separation by a graph problem



Hella et al. (PODC 2012):

- Logical characterisations for constant-time variants of the problem classes
- In a certain class of structures, SV corresponds to *multimodal logic*
- ... and MV corresponds to graded multimodal logic.

Hella et al. (PODC 2012):

- Logical characterisations for constant-time variants of the problem classes
- In a certain class of structures, SV corresponds to *multimodal logic*
- ... and MV corresponds to graded multimodal logic.
- Our result: When given a formula ϕ of graded multimodal logic, we can find an equivalent formula ψ of multimodal logic, but in general, the modal depth md(ψ) of ψ has to be at least md(ϕ) + Δ 1.

Conclusion

 $\mathcal{MV}:$ Send a vector, receive a multiset.

SV: Send a vector, receive a set.

Previously:

• It is possible to simulate \mathcal{MV} in \mathcal{SV} by using $2\Delta - 2$ extra rounds.

This work:

- $2\Delta 2$ rounds are necessary to solve the simulation problem.
- There is a graph problem for which the difference in running time between \mathcal{MV} and \mathcal{SV} is $\Delta 1$ rounds.

The thesis A Classification of Weak Models of Distributed Computing is available at http://hdl.handle.net/10138/144214.