

Node Labels in Local Decision

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Abstract. The role of unique node identifiers in network computing is well understood as far as *symmetry breaking* is concerned. However, the unique identifiers also *leak information* about the computing environment—in particular, they provide some nodes with information related to the size of the network. It was recently proved that in the context of *local decision*, there are some decision problems such that (1) they cannot be solved without unique identifiers, and (2) unique node identifiers leak a *sufficient* amount of information such that the problem becomes solvable (Fraigniaud et al., PODC 2013).

In this work we study what is the *minimal* amount of information that we need to leak from the environment to the nodes in order to solve local decision problems. Our key results are related to *scalar oracles* f that, for any given n , provide a multiset $f(n)$ of n labels; then the adversary assigns the labels to the n nodes in the network. This is a direct generalisation of the usual assumption of unique node identifiers. We give a complete characterisation of the *weakest oracle* that leaks at least as much information as the unique identifiers.

Our main result is the following dichotomy: we classify scalar oracles as *large* and *small*, depending on their asymptotic behaviour, and show that (1) any large oracle is at least as powerful as the unique identifiers in the context of local decision problems, while (2) for any small oracle there are local decision problems that still benefit from unique identifiers.

Keywords: distributed computing, local decision

1 Introduction

This work studies the role of *unique node identifiers* in the context of *local decision problems* in distributed systems. We generalise the concept of node identifiers by introducing *scalar oracles* that choose the labels of the nodes, depending on the size of the network n —in essence, we let the oracle leak some information on n to the nodes—and ask what is the *weakest* scalar oracle that we could use instead of unique identifiers (see Figure 1). We prove the following dichotomy: we classify each scalar oracle as *small* or *large*, depending on its asymptotic behaviour, and we show that the large oracles are precisely those oracles that are at least as strong as unique identifiers.

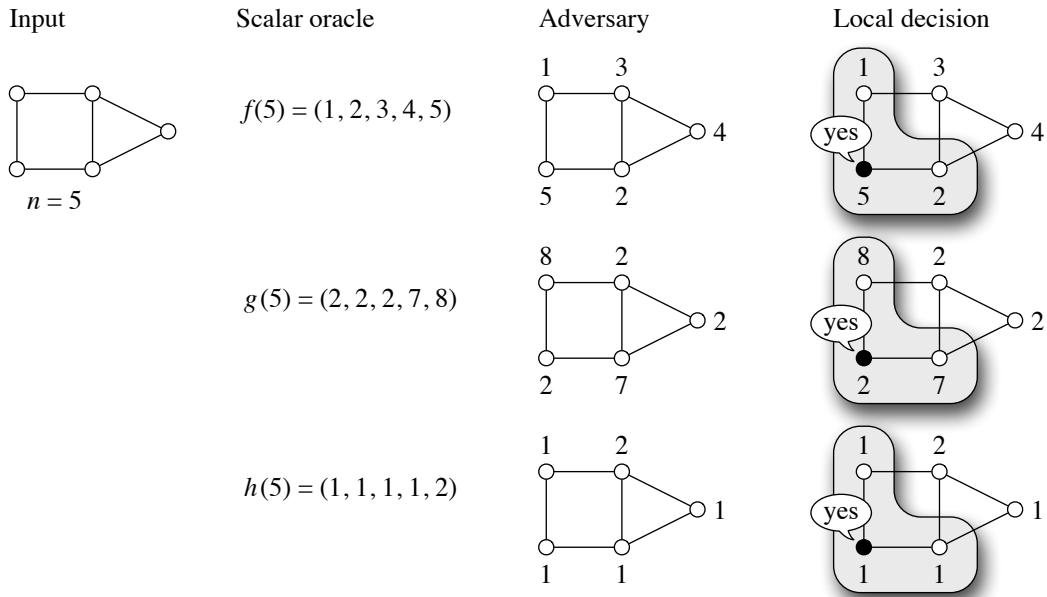


Figure 1: A scalar oracle produces a list of n labels, the adversary assigns the labels to the nodes, and a local decider accepts or rejects the input based on the local neighbourhoods of the nodes. Here f and g are examples of *large* scalar oracles, while h is a *small* scalar oracle—informally, f and g ensure that the maximum of any subset of k labels grows with k , while h violates this. Scalar oracle f produces unique identifiers, while g and h may produce duplicate labels. There are many problems that become more difficult to solve with distributed algorithms if there are duplicate labels. Nevertheless, we will see that in the context of *local decision*, any large scalar oracle is at least as informative as f , while this is not the case for any small scalar oracle.

1.1 Context and background

The research trends within the framework of distributed computing are most often pragmatic. Problems closely related to real world applications are tackled under computational assumptions reflecting existing systems, or systems whose future existence is plausible. Unfortunately, small variations in the model settings may lead to huge gaps in terms of computational power. Typically, some problems are unsolvable in one model but may well be efficiently solvable in a slight variant of that model. In the context of *network computing*, this commonly happens depending on whether the model assumes that pairwise distinct identifiers are assigned to the nodes. While the presence of distinct identifiers is inherent to some systems (typically, those composed of artificial devices), the presence of such identifiers is questionable in others (typically, those composed of biological or chemical elements). Even if the identifiers are present, they may not necessarily be directly visible, e.g., for privacy reasons.

The absence of identifiers, or the difficulty of accessing the identifiers, limits the power of computation. Indeed, it is known that the presence of identifiers ensures two crucial properties, which are both used in the design of efficient algorithms. One such property is **symmetry breaking**. The absence of identifiers makes symmetry breaking more difficult to achieve, or even impossible if asymmetry cannot be extracted from the inputs of the nodes, from the structure of the network, or from some source of random bits. The role of the identifiers in the framework of network computing, as far as symmetry breaking is concerned, has been investigated in depth, and is now well understood [1–9, 16–24, 27–29].

The other crucial property of the identifiers is their ability to **leak global information** about the framework in which the computation takes place. In particular, the presence of pairwise distinct identifiers guarantees that at least one node has an identifier at least n in n -node networks. This apparently very weak property was proven to actually play an important role when one is interested in checking the correctness of a system configuration in a decentralised manner. Indeed, it was shown in prior work [13] that the ability to check the legality of a system configuration with respect to some given Boolean predicate differs significantly according to the ability of the nodes to use their identifiers. This phenomenon is of a nature different from symmetry breaking, and is far less understood than the latter.

More precisely, let us define a *distributed language* as a set of system configurations (e.g., the set of properly coloured networks, or the set of networks each with a unique leader). Then let LD be the class of distributed languages that are *locally decidable*. That is, LD is the set of distributed languages for which there exists a distributed algorithm where every node inspects its neighbourhood at constant distance in the network, and outputs *yes* or *no* according to the following rule: all nodes output *yes* if and only if the instance is legal. Equivalently, the instance is illegal if and only if at least one node outputs *no*. Let LDO be defined as LD with the restriction that the local algorithm is required to be *identifier oblivious*, that is, the output of every node is the same regardless of the identifiers assigned to the nodes. By definition, $\text{LDO} \subseteq \text{LD}$, but [13] proved that this inclusion is strict: there are languages in $\text{LD} \setminus \text{LDO}$. This strict inclusion was obtained by constructing a distributed language that can be decided by an algorithm whose outputs depend heavily on the identifiers assigned to the nodes, and in particular on the fact that at least one node has an identifier whose value is at least n .

The gap between LD and LDO has little to do with symmetry breaking. Indeed, decision tasks do not require that some nodes act differently from the others: on legal instances, all nodes must output *yes*, while on illegal instances, it is permitted (but not required) that all nodes output *no*. The gap between LD and LDO is entirely due to the fact that the identifiers leak information about the size n of the network. Moreover, it is known that the gap between LD and LDO is strongly related to computability issues: there is an identifier-oblivious *non-computable* simulation A' of every local algorithm A that uses identifiers to decide a distributed language [13]. Informally, this means that for every language in $\text{LD} \setminus \text{LDO}$, unique identifiers are as helpful as providing the nodes with the capability of solving the halting problem.

1.2 Objective

One objective of this paper is to measure the *amount of information* provided to a distributed system via the labels given to its nodes. For this purpose, we consider the classes LD and LDO enhanced with *oracles*, where an oracle f is a function that provides every node with information about its environment.

We focus on the class of *scalar* oracles, which are functions over the positive integers. Given an $n \geq 1$, a scalar oracle f returns a list $f(n) = (f_1, f_2, \dots, f_n)$ of n labels (bit strings) that are assigned arbitrarily to the nodes of any n -node network in a one-to-one manner. The class LD^f (resp., LDO^f) is then defined as the class of distributed languages decidable locally by an algorithm (resp., by an identifier-oblivious algorithm) in networks labelled with oracle f .

If, for every $n \geq 1$, the n values in the list $f(n)$ are pairwise distinct, then $\text{LD} \subseteq \text{LDO}^f$ since the nodes can use the values provided to them by the oracle as identifiers. However, as we shall demonstrate in the paper, this pairwise distinctness condition is not necessary.

Our goal is to identify the interplay between the classes LD , LDO , LD^f , and LDO^f , with respect to any scalar oracle f , and to characterise the power of identifiers in distributed systems as far as leaking information about the environment is concerned.

1.3 Our results

Our first result is a characterisation of the weakest oracles providing the same power as unique node identifiers. We say that a scalar oracle f is *large* if, roughly, f ensures that, for any set of k nodes, the largest value provided by f to the nodes in this set grows with k (see Section 2.3 for the precise definition). We show the following theorem.

Theorem 1. *For any computable scalar oracle f , we have $\text{LDO}^f = \text{LD}^f$ if and only if f is large.*

Theorem 1 is a consequence of the following two lemmas. The first says that small oracles (i.e. non-large oracles) do not capture the power of unique identifiers. Note that the following separation result holds for any small oracle, including uncomputable oracles.

Lemma 2. *For any small oracle f , there exists a language $L \in \text{LD} \setminus \text{LDO}^f$.*

The second is a simulation result, showing that any local decision algorithm using identifiers can be simulated by an identifier-oblivious algorithm with the help of *any* large oracle, as long as the oracle itself is computable. Essentially large oracles capture the power of unique identifiers.

Lemma 3. *For any large computable oracle f , we have $\text{LD} \subseteq \text{LDO}^f = \text{LD}^f$.*

Theorem 1 holds despite the fact that small oracles can still produce some large values, and that there exist small oracles guaranteeing that, in any n -node network, at least one node has a value at least n . Such a small oracle would be sufficient to decide the language $L \in \text{LD} \setminus \text{LDO}$ presented in [13]. However, it is not sufficient to decide all languages in LD .

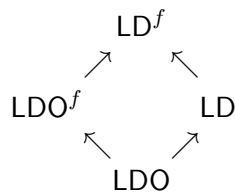
Our second result is a complete description of the hierarchy of the four classes LD , LDO , LD^f , and LDO^f of local decision, using identifiers or not, with or without oracles. The pictures for small and large oracles are radically different.

- For any large oracle f , the hierarchy yields a *total order*:

$$\text{LDO} \subsetneq \text{LD} \subseteq \text{LDO}^f = \text{LD}^f.$$

The strict inclusion $\text{LDO} \subsetneq \text{LD}$ follows from [13]. The second inclusion $\text{LD} \subseteq \text{LDO}^f$ may or may not be strict depending on oracle f .

- For a small oracle f , the hierarchy yields a *partial order*. We have $\text{LDO}^f \subsetneq \text{LD}^f$ as a consequence of Lemma 2. However, LD and LDO^f are incomparable, in the sense that there is a language $L \in \text{LD} \setminus \text{LDO}^f$ for any small oracle f , and there is a language $L \in \text{LDO}^f \setminus \text{LD}$ for some small oracles f . Hence, in the general case, the relationships of the four classes can be represented as the following diagram:



All inclusions (represented by arrows) are strict at least for some choice of f . Note that it is also possible to choose an unhelpful small oracle f such that the hierarchy collapses to $\text{LDO} = \text{LDO}^f \subsetneq \text{LD} = \text{LD}^f$.

1.4 Additional related work

In the context of network computing, oracles and advice commonly appear in the form of *labelling schemes* [10, 15]. A typical example is a *distance labelling scheme*, which is a labelling of the nodes so that the distance between any pair of nodes can be computed or approximated based on the labels. Other examples are *routing schemes* that label the nodes with information that helps in finding a short path between any given source and destination. For graph problems, one could of course encode the entire solution in the advice string—hence the key question is whether a very small amount of advice helps with solving a given problem.

In prior work, it is commonly assumed that the oracle can give a specific piece of advice for each individual node. The advice is localised, and entirely controlled by the oracle. Moreover, the oracle can see the entire problem instance and it can tailor the advice for any given task.

In the present work, we study a much weaker setting: the oracle is only given n , and it cannot choose which label goes to which node. This is a generalisation of, among others, typical models of *networks with unique identifiers*: one commonly assumes that the unique identifiers are a permutation of $\{1, 2, \dots, n\}$ [21], which in our case is exactly captured by the large scalar oracle

$$f(n) = (1, 2, \dots, n),$$

or that the unique identifiers are a subset of $\{1, 2, \dots, n^c\}$ for some constant c [26], which in our case is captured by a subfamily of large scalar oracles. Our model is also a generalisation of *anonymous networks with a unique leader* [9]—the assumption that there is a unique leader is captured by the small scalar oracle

$$f(n) = (0, 0, \dots, 0, 1).$$

2 Model and definitions

In this work, we augment the usual definitions of *locally checkable labellings* [23] and *local distributed decision* [11–13] with scalar oracles.

2.1 Computational model

We deal with the standard LOCAL model [26] for distributed graph algorithms. In this model, the network is a simple connected graph $G = (V, E)$. Each node $v \in V$ has an *identifier* $\text{id}(v) \in \mathbb{N}$, and all identifiers of the nodes in the network are pairwise distinct. Computation proceeds in synchronous rounds. During a round, each node communicates with its neighbours in the graph, and performs some local computation. There are no limits on the amount of communication done in a single round. Hence, in r communication rounds, each node can learn the complete topology of its radius- r neighbourhood, including the inputs and the identifiers of the nodes in this neighbourhood. In a distributed algorithm, all nodes start at the same time, and each node must halt after some number of rounds, and produce its individual output. The collection of individual outputs then forms the global output of the computation. The running time of the algorithm is the number of communication rounds until all nodes have halted.

We consider *local* algorithms, i.e., constant-time algorithms [27]. That is, we focus on algorithms with a running time that does not depend on the size n of the graph. Any such algorithm, with a running time of r , can be seen as a function from the set of all possible radius- r neighbourhoods to the set of all possible outputs.

An *identifier-oblivious* algorithm is an algorithm whose output is independent of the identifiers assigned to the nodes. Note that, from the perspective of an identifier-oblivious algorithm, the set of all possible radius- r degree- d neighbourhoods is finite. This is not the case for every algorithm since there are infinitely many identifier assignments to the nodes in a radius- r degree- d neighbourhood.

Although the LOCAL model does not put any restriction on the amount of individual computation performed at each node, we only consider algorithms that are *computable*.

2.2 Local decision tasks

We are interested in the power of constant-time algorithms for *local decision*. A *labelled graph* is a pair (G, x) , where G is a simple connected graph, and $x: V(G) \rightarrow \{0, 1\}^*$ is a function assigning a label to each node of G . A *distributed language* L is a set of labelled graphs. Examples of distributed languages include:

- 2-colouring, the language where G is a bipartite graph and we have $x(v) \in \{0, 1\}$ for each $v \in V(G)$ such that $x(v) \neq x(u)$ whenever $\{u, v\} \in E(G)$;
- parity, the language of graphs with an even number of nodes;
- planarity, the language that consists of all planar graphs.

We say that algorithm A decides L if and only if the output of A at every node is either *yes* or *no*, and, for every instance (G, x) , algorithm A satisfies:

$$(G, x) \in L \iff \text{all nodes output } \textit{yes}.$$

Hence, for an instance $(G, x) \notin L$, the algorithm A must ensure that at least one node outputs *no*. We consider two main distributed complexity classes:

- LD (for *local decision*) is the set of languages decidable by constant-time algorithms in the LOCAL model.
- LDO (for *local decision oblivious*) is the set of languages decidable by constant-time identifier-oblivious algorithms in the LOCAL model.

By definition, $\text{LDO} \subseteq \text{LD}$, and it is known [13] that this inclusion is strict: there are languages $L \in \text{LD} \setminus \text{LDO}$. The fact that we consider only computable algorithms is crucial here—without this restriction we would have $\text{LDO} = \text{LD}$ [13].

2.3 Distributed oracles

We study the relationship of classes LD and LDO with respect to *scalar oracles*. Such an oracle f is a function that assigns a list of n values to every positive integer n , i.e.,

$$f(n) = (f_1, f_2, \dots, f_n)$$

with $f_i \in \mathbb{N}$. In essence, oracle f can provide some information related to n to the nodes. In an n -node graph, each of the n nodes will receive a value $f_i \in f(n)$, $i \in [n]$. These values are arbitrarily assigned to the nodes in a one-to-one manner. Two different nodes will thus receive f_i and f_j with $i \neq j$. Note that f_i may or may not be different from f_j for $i \neq j$; this is up to the choice of the oracle. The way the values provided by the oracles are assigned to the nodes is under the control of an adversary. One example of an oracle is $f(n) = (1, 2, \dots, n)$, which provides the nodes with identifiers. Another example is $f(n) = (0, 0, \dots, 0)$, which provides no information to the nodes.

W.l.o.g., let us assume that $f_i \leq f_{i+1}$ for every i . We use the shorthand $f_k^{(n)}$ for the k th label provided by f on input n , that is, $f(n) = (f_1^{(n)}, f_2^{(n)}, \dots, f_n^{(n)})$. For a fixed oracle f , we consider two main distributed complexity classes:

- LD^f is the set of languages decidable by constant-time algorithms in networks that are labelled with oracle f .
- LDO^f is the set of languages decidable by constant-time identifier-oblivious algorithms in networks that are labelled with oracle f .

We will separate oracles in two classes, which play a crucial role in the way the four classes LDO, LD, LDO^f, and LD^f interact.

Definition 1. *An oracle f is said to be large if*

$$\forall c > 0, \exists k \geq 1, \forall n \geq k, f_k^{(n)} \geq c.$$

An oracle is small if it is not large.

Since we have by definition that $f_m^{(n)} \geq f_k^{(n)}$ for all $m > k$, a large oracle f satisfies that for any value $c > 0$, there exists a large enough k , such that, in every graph G of size at least k , for every set of nodes $S \subseteq V(G)$ of size $|S| \geq k$, oracle f is providing at least one node of S with a value at least as large as c . In short: every large set of nodes must include at least one node that receives a large value.

Conversely, a small oracle f satisfies that there exists a value $c > 0$ such that, for every k , we can find $n \geq k$ such that, in every n -node graph G , and for every set of nodes $S \subseteq V(G)$ of size $|S| \geq k$, there is an assignment of the values provided by f such that every node in S receives a value smaller than c . In short: there are arbitrarily large sets of nodes which all receive a small value.

For example, oracles $f(n) = (1, 2, \dots, n)$ and $f(n) = (n, n, \dots, n)$ are large, while oracles $f(n) = (0, 0, \dots, 0, 1)$ and $f(n) = (0, 0, \dots, 0, 2^n)$ are small. We emphasise that small oracles can output very large values, and large oracles can output small values: For example, the values of a large oracle f with $f_i^{(n)} = \log^* n$ are at most 5 for $n \leq 2^{65536}$. On the other hand, the oracle g , with $g_i^{(n)} = 1$ if $i \leq \log^* n$, and $g_i^{(n)} = 2^i$ otherwise, is small.

Many natural processes for node labelling correspond to small oracles. Consider, for example, an oracle that labels a fraction 1/2 of nodes with label 1, a fraction 1/4 of nodes with label 2, a fraction 1/8 of nodes with label 3, etc. Here the range of labels increases with n , but nevertheless the oracle is small, and our results imply that it is strictly weaker than any large oracle.

3 Proof of the main theorem

In this section we give the proof of our main result that characterises the power of small and large oracles with respect to identifier-oblivious local decision.

3.1 Small oracles do not capture the power of unique identifiers

In this section, we will prove the following lemma:

Lemma 2. *For any small oracle f , there exists a language $L \in \text{LD} \setminus \text{LDO}^f$.*

Fraigniaud et al. [13] showed that there exists a language $L \in \text{LD} \setminus \text{LDO}$. We use a Turing machine construction similar to the one in their proof. However, we must take into account the additional concern of the values that the oracle assigns to the nodes. The key idea is that our construction consists of two parts, a *body* and a *tail*; here the body is directly borrowed from Fraigniaud et al. [13] while the tail is simply a sufficiently long path. We will then construct a language L that is solvable if and only if at least one node in the body has access to some information on the size of the body. On the one hand, we will have $L \in \text{LD}$, as a large body will imply that some nodes in the body have large identifiers. On the other hand, we will have $L \notin \text{LDO}^f$, as any small oracle f can output many “unhelpful” labels that do not leak any information on the size of the network, and the adversary may choose an assignment in which the nodes in the body will only see unhelpful labels.

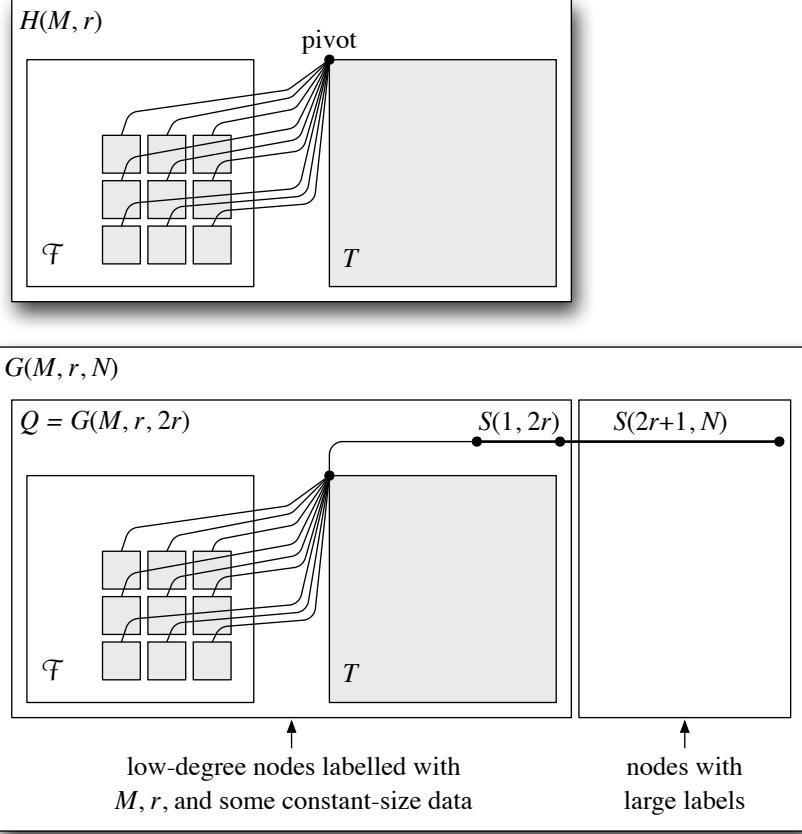


Figure 2: The construction of graph $G(M, r, N)$ in our proof, and a comparison with graph $H(M, r)$ from Fraigniaud et al. [13]. Our construction simply augments $H(M, r)$ with a path of length N .

Proof overview. For each Turing machine M that halts on the empty tape, for each locality parameter $r \in \mathbb{N}$, and for each size parameter $N \in \mathbb{N}$, we will define a labelled graph $G(M, r, N)$, which consists of two parts, a body and a tail. The body encodes the execution table of M and is completely determined by the parameters M and r , and the tail is simply a path of N nodes. See Figure 2 for illustrations.

We also define the following languages related to recognition of graphs from family $G(M, r, N)$:

$$L_{\text{structure}} = \{G(M, r, N) : r \geq 1, N \geq 1\},$$

$$L_{\text{output}} = \{G(M, r, N) : r \geq 1, N \geq 1, \text{ and Turing machine } M \text{ outputs } 0\}.$$

We will then show that $L_{\text{output}} \in \text{LD}$ and $L_{\text{output}} \notin \text{LDO}^f$ for any small oracle f , establishing Lemma 2.

Properties of the construction. Our construction $G(M, r, N)$ is designed so that it will satisfy the following properties:

- (P1) Each label of the body of $G(M, r, N)$ is a triple $x(v) = (M, r, x'(v))$ for some $x'(v)$ which is bounded by a computable function of M and r .
- (P2) The labels of the tail of $G(M, r, N)$ only depend on N . In particular, they do not carry any information on M .
- (P3) The maximum degree of $G(M, r, N)$ is bounded by a computable function of M and r .

- (P4) The number of nodes in the body of $G(M, r, N)$ is at least as large as the number of steps M takes on an empty tape. In particular, if there exists a finite $G(M, r, N)$, then M halts.
- (P5) The number of nodes in the tail of $G(M, r, N)$ is N .
- (P6) We have $L_{\text{structure}} \in \text{LDO}$.
- (P7) We have $L_{\text{output}} \notin \text{LDO}^f$ for any small oracle f .

We will first construct $G(M, r, N)$ incrementally, carefully maintaining properties (P1)–(P3) and working towards a construction that also satisfies properties (P4)–(P7). It is easy to see that the construction indeed satisfies properties (P1)–(P5), and we will later argue that the construction also satisfies properties (P6) and (P7). Finally, we show how Lemma 2 follows directly from these properties.

Construction for (P4). Assume that M halts in s rounds on the empty input. The first part of the construction $G(M, r, N)$ is a grid graph T of size $(s + 1) \times (s + 1)$, called the *execution table* of M . We denote each node by a pair (x, y) , where x is the row and y is the column. Two nodes are adjacent if their Euclidean distance is one. The execution table is encoded in a standard way, with each row x representing the tape of the Turing machine after step x and each node (x, y) holding the contents of the tape in cell y . Edges are given an orientation, from left to right and from top to bottom, using labels $(x \bmod 3, y \bmod 3)$ for each node (x, y) . It is essential that we do not give nodes the full label (x, y) , which would leak information on s and violate property (P1).

On each row, there is a single node that holds the machine head for M . This node also knows the state of M . On the row 0, this has to be the node $(0, 0)$, known as the *pivot*.

We now have a construction that satisfies property (P4), but not necessarily the other properties; the structure is not locally checkable, the size of T is not N , and all nodes that see the state of M on the last row know whether M halts with 0 or 1.

Construction for (P5). To satisfy property (P5), we augment the construction with a *tail* that consists of a path $S_N = (s_1, s_2, \dots, s_N)$. The first node s_1 is connected to the pivot and each node s_j is labelled with j , which is its distance to the pivot; note that this is compatible with property (P2). In the analysis, we will denote any subsequence $(s_i, s_{i+1}, \dots, s_j)$ of the tail by $S(i, j)$.

Construction for (P6). We augment T with a pyramidal structure, a quadtree of height $h = \log_2(s + 1)$; see Figure 3. Here we will assume that $s + 1$ is a power of 2—if this is not the case, we can pad each row with empty cells and add rows that are copies of the row with the stopping state.

In the quadtree, each level z consists of a square grid $[(s + 1)/2^z] \times [(s + 1)/2^z] \times \{z\}$, with level 0 equal to T . Again, nodes on each level z are connected if their Euclidean distance is at most one. Each node (x, y, z) on level $z < h$ is connected to node $(\lceil x/2 \rceil, \lceil y/2 \rceil, z + 1)$. Nodes on levels $z > 0$ only get (M, r) as their label. Denote T augmented with the quadtree by T^Δ .

The quadtree structure is essential in order to satisfy property (P6). We will revisit this property soon and argue that the structure is indeed locally checkable, but we will first describe the remaining parts of the construction.

Construction for (P7). Currently the final output of Turing machine M is directly visible at the bottom row of the execution table, violating property (P7). To fix that, a collection of all syntactically possible local neighbourhoods is attached to the construction. This is done so that if the local decider only tries to locally scan the graph, it will always find a local neighbourhood

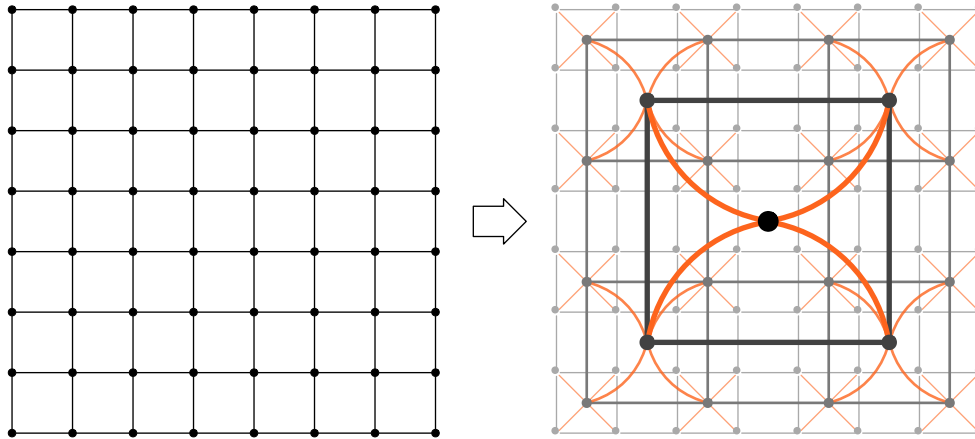


Figure 3: Construction of T^Δ : augmenting grid graph T with a quadtrees structure.

of any type that it is looking for. We will later formalize this intuition by showing that the property of having every possible local neighbourhood consistent with M and r will allow a sequential algorithm to simulate any identifier-oblivious algorithm.

Observe that once we fix a Turing machine M , we also fix the set of states and the alphabet, and hence also the set of possible labels that may occur in the execution table T . Thus for any M there are finitely many different neighbourhoods that may appear in T .

We construct a *fragment collection* $\mathcal{F} = \mathcal{F}(M, r)$ of $(2^{3r} \times 2^{3r})$ grid fragments. There will be a fragment $F \in \mathcal{F}$ for each possible subgrid that could appear on T according to the transition rules of M : every 2×2 subgrid must be consistent with the state transitions of M , and the modulo-3 labels provide a consistent orientation. Both properties can be checked locally. Each fragment F is augmented with a quadtrees F^Δ of height $3r$, as in the construction of T^Δ , and each node in the tree is given the universal label (M, r) . We denote this collection of fragments by $\mathcal{F}^\Delta = \mathcal{F}^\Delta(M, r)$. The height $3r$ is chosen to ensure that every possible r -neighbourhood that could syntactically appear in T^Δ will appear in \mathcal{F}^Δ .

Since, as we observed, the labellings are of constant size, also when accounting for the quad tree, and the fragments are of constant size, there is a computable sequential algorithm for generating the fragment collection \mathcal{F}^Δ , and the algorithm halts even if M does not. We will use this observation later to show that property (P7) is indeed satisfied.

Next, the fragment collection needs to be made connected and locally checkable. The pivot will be connected to every fragment F^Δ and will be responsible for checking that \mathcal{F}^Δ indeed contains every possible fragment. To this end, we define the *natural* borders of a fragment F as follows.

- The top row of F is never a natural border.
- The bottom row of F is natural if the machine head does not appear on it in a non-halting state.
- The leftmost column of F is a natural border if it could on a syntactical basis appear as the leftmost column of T , that is, if the machine head of M never appears from or disappears to the left of that column.
- The rightmost column of F is a natural border under analogous circumstances.
- However, if both the leftmost and the rightmost column of F would be natural, replace the fragment with two copies in which the leftmost and the rightmost column, in turn, are considered non-natural.

We connect every node of each non-natural border of every fragment to the pivot. Since all labellings are of constant size, and there is constant number of different fragments in \mathcal{F} , each node, including the pivot, has constant maximum degree, as required by property (P3).

Observe that in each fragment, the top row and at least one of the sides is non-natural. This *border property* ensures that each possible fragment $F^\Delta \in \mathcal{F}^\Delta$ is uniquely reconstructed, using the transition rules of M , based on the non-natural borders. To see this, note that the top row fixes the initial contents of each tape cell, and the non-natural borders fix whether the machine head is present in the fragment. The border property also ensures that the non-natural borders induce a connected subgraph of the fragment.

We have now described all parts of the construction; however, it is not obvious that properties (P6) and (P7) indeed hold. We will next prove that this is the case.

Property (P6). We will show that $L_{\text{structure}} \in \text{LDO}$, i.e., the task of recognising graphs from the family $G(M, r, N)$ is in LDO. To recognise that the input graph G is equal to some $G(M, r, N)$, we proceed according to the following steps, with nodes rejecting if any property is violated.

Step 1: Nodes outside the tail make sure that the label (M, r) is consistent throughout G .

Step 2: Each node of G should be either part of a quadtree or in the tail.

Quadtree: Nodes can distinguish between edges that are contained within one level of the quadtree, edges that connect different levels of the quadtree (*cross-level edges*), and edges that connect a node to the pivot (*cross-pyramid edges*). Nodes can also tell which one, if any, of the cross-level edges goes up to a higher level. By construction, there must be a single node at the top of the pyramid. This node fixes the structure of the four nodes below it, making sure that they form a 4-cycle. These in turn fix the boundaries of each square grid, making it possible to check that none of the grids are wrapped around as toruses or deviate from the construction in any other way. Thus nodes can check that, apart from the tail, G consists of square execution table grids and edges connecting them. Inside each bottom layer, nodes can check that the orientation is consistent.

Tail: Nodes check that the distance counter in the tail is consistent, and that the tail is indeed a path rooted at the pivot.

Step 3: Inside each square grid at the bottom layer, nodes can check that every (2×2) subgrid is consistent with the state transitions of M . This ensures that the execution table and every fragment represent a possible execution of M . In particular, on T^Δ this ensures that the encoded execution is that of M on the empty tape.

Step 4: Nodes at the borders of the bottom layer grids fall into three cases.

- (1) Pivot: This is a node incident to multiple cross-pyramid edges. Check that the node is located at the top left corner of a grid. Check that all cross-pyramid edges are connected to non-natural borders of fragments. Check that the border property is satisfied for each fragment, and the non-natural borders form connected subgraphs.
- (2) Non-natural borders: These are nodes incident to exactly one cross-pyramid edge. Check that the cross-pyramid edge is connected to a pivot. Check that the adjacent nodes follow the border property. Check that adjacent non-natural border nodes are connected to the *same* pivot—this ensures that the pivot is unique.
- (3) Other borders: These are nodes without cross-pyramid edges. Check that the machine head does not cross the border, i.e., we have a natural border or the top row of the execution table.

Step 5: Finally, the pivot can construct the set of fragments \mathcal{F} and, using the border property, check that every fragment is present exactly once.

Property (P7). Let f be any small oracle. We show that L_{output} is not in LDO^f . For the sake of contradiction, we assume that there is an LDO^f -algorithm A that decides L_{output} . It is known that separating the following languages is undecidable (see e.g. [25, p. 65]):

$$L_i = \{M : \text{Turing machine } M \text{ outputs } i\} : i \in \{0, 1\}. \quad (1)$$

We will use algorithm A to construct a sequential algorithm B that separates L_0 and L_1 . As such a B cannot exist, A cannot exist either; it follows that L_{output} is not in LDO^f .

We start with some preliminary observations. Let r be the running time of A , and consider the execution of A on an instance $G(M, r, N)$ for some M and N . It follows that each node in $S(r+1, N) \subseteq G(M, r, N)$ must always output *yes*. To see this, note that the claim is trivial if M halts with 0. Otherwise we can always construct another instance $G(M_0, r, N)$ such that M_0 halts with 0 and both $G(M, r, N)$ and $G(M_0, r, N)$ have the same number of nodes. Hence the oracle and the adversary can assign the same labels to $S(r+1, N)$ in both $G(M, r, N)$ and $G(M_0, r, N)$. If any of these nodes would answer *no* in $G(M, r, N)$, then A would also incorrectly reject the *yes*-instance $G(M_0, r, N) \in L_{\text{output}}$.

We will now choose an appropriate value of N as follows. Let b denote the number of nodes in the body of $G(M, r, \cdot)$; that is, there are $b + N$ nodes in $G(M, r, N)$. By the definition of a small oracle, there exists a c such that for all k oracle $f(n)$ outputs some label $i \in [c]$ at least $\lceil k/c \rceil$ times on some $n \geq k$. Moreover, we can find an infinite sequence of values k_0, k_1, \dots such that the most common value is some fixed i_0 . We select w.l.o.g. the smallest k_j and a suitable n such that $f(n)$ contains at least $k_j/c \geq b + 2r$ labels equal to i_0 . Let $N = n - b$. With such a choice of N , our adversary can produce the following *worst-case labelling*: all nodes of $G(M, r, 2r)$ receive the constant input label i_0 , and all other labels are assigned to the nodes in $S(2r+1, N)$.

Note that $G(M, r, 2r)$ is equal to the union of the radius- r neighbourhoods of the nodes of $G(M, r, r)$. To recap, so far we have achieved the following:

1. Only nodes in $G(M, r, r)$ produce useful output—all other nodes always output *yes*.
2. The nodes of $G(M, r, r)$ see only the “unhelpful” oracle label i_0 in their radius- r neighbourhoods. All other oracle labels are beyond the reach of these nodes in algorithm A .

Furthermore, note that the constant i_0 only depends on the choice of f .

Now given a Turing machine M , algorithm B proceeds as follows. Consider the subgraph $Q = G(M, r, 2r) \subseteq G(M, r, N)$, and assume the worst-case labelling of $G(M, r, N)$ in which all nodes of Q have the constant label i_0 . Algorithm B cannot construct Q ; indeed, M might not halt, in which case $G(M, r, N)$ would not even exist. However, B can do the following: it can assume that M halts, and then generate a collection \mathcal{Q} that would contain all possible radius- r neighbourhoods of the nodes in $G(M, r, r)$. Since the top rows do not appear as natural in \mathcal{F} , collection \mathcal{Q} consists of all neighbourhoods that intersect with the top row of T^Δ , and every r -neighbourhood appearing in \mathcal{F}^Δ . This collection is finite and can be, as observed previously, generated using a computable algorithm.

Next B will simulate A in each neighbourhood of \mathcal{Q} . If M halts with 1, then $G(M, r, N) \notin L_{\text{output}}$, and therefore one of the nodes in $G(M, r, r)$ has to output *no*; in this case B outputs 1. If M halts with 0, then $G(M, r, N) \in L_{\text{output}}$, and therefore all nodes in $G(M, r, r)$ have to output *yes*; in this case B outputs 0. The key observation is that B will always halt with some (meaningless) output even if we are given an input $M \notin L_0 \cup L_1$; hence B is a computable function that separates L_0 and L_1 . As such a B cannot exist, A cannot exist either.

Completing the proof of Lemma 2. We have already established that $L_{\text{output}} \notin \text{LDO}^f$ for any small f . Also, $L_{\text{structure}} \in \text{LDO}$, which trivially implies $L_{\text{structure}} \in \text{LD}$. To complete the proof, it suffices to show that we also have $L_{\text{output}} \in \text{LD}$.

The LD-algorithm for L_{output} proceeds as follows. Given an input graph G , we first verify that $G \in L_{\text{structure}}$. If this indeed holds, we have $G = G(M, r, N)$ for some Turing machine M , and all nodes in the body know M . Now each node v in the body simulates M for $\text{id}(v)$ steps. If M halts with 1 during the simulation, v rejects the input, otherwise it accepts the input.

The algorithm clearly rejects any $G \notin L_{\text{structure}}$, and it clearly accepts any $G \in L_{\text{output}}$. To see that it rejects any $G \in L_{\text{structure}} \setminus L_{\text{output}}$, observe that by property (P4) the number of nodes in the body of $G(M, r, N)$ is at least as large as the running time of M . Hence at least one node of the body has an identifier at least as large as the running time of M , and if M halts with 1, it will be detected. \square

3.2 Large oracles capture the power of unique identifiers

In this section we will show that a *computable* large oracle f is sufficient to have $\text{LD} \subseteq \text{LDO}^f = \text{LD}^f$. This result holds even if f only has access to an upper bound $N \geq n$, and the adversary gets to pick an n -subset of labels from $f(N)$. Note that the oracle has to be computable in order for us to invert it locally.

Lemma 3. *For any large computable oracle f , we have $\text{LD} \subseteq \text{LDO}^f = \text{LD}^f$.*

Proof. We begin by showing how to recover an oracle \hat{f} with $\hat{f}_k^{(N)} \geq k$, for all k and $N \geq k$, from a large oracle f . We want to guarantee that each node v receives a label $\ell \geq i$ if in the initial labelling it had the i th smallest label.

By definition, it holds for large oracles that for each natural number ℓ there is a largest index i such that $f_i^{(N)} \leq \ell$; we denote the index by $g(\ell)$. By assumption, a node with label ℓ can locally compute the value $g(\ell)$. We now claim that

$$\hat{f}: N \mapsto \{g(f_1), g(f_2), \dots, g(f_N)\}$$

has the property $\hat{f}_k^{(N)} \geq k$. To see this, assume that we have $f_k^{(N)} = \ell$ for an arbitrary k . Seeing label ℓ , node v knows that, in the worst case, its own label is the $g(\ell)$ th smallest. Thus for every k , the node with the k th smallest label will compute a new label at least k .

Now given \hat{f} , we can simulate any r -round LD-algorithm A as follows.

1. Each node v with label ℓ_v locally computes the new label $g(\ell_v)$.
2. Each node gathers all labels $g(\ell_u)$ in its r -neighbourhood. Denote by g_v^* the maximum value in the neighbourhood of v .
3. Each node v simulates A on every unique identifier assignment to its local r -neighbourhood from $\{1, 2, \dots, g_v^*\}$. If for some assignment A outputs *no*, then v outputs *no*, and otherwise it outputs *yes*.

Because of how the decision problem is defined, it is always safe to output *no* when some simulation of A outputs *no*. It remains to be argued that it is safe to say *yes*, if all simulations say *yes*. This requires that *some* subset of simulations of A , one for each node, looks as if there had been a consistent setting of unique identifiers on the graph. Now let id be one identifier assignment with $\text{id}(v) = i$ for the v with i th smallest label, for all i (breaking ties arbitrarily). Since by construction $g(\ell_v) \geq \text{id}(v)$ for all v , there will be a simulation of A for every node v with local identifier assignment id_v such that for all u in the radius- r neighbourhood of v we have $\text{id}_v(u) = \text{id}(u)$.

So far we have seen how to simulate any LD-algorithm A with LDO^f -algorithms. We can apply the same reasoning to simulate any LD^f -algorithm A with LDO^f -algorithms; the only difference is that each node in the simulation has now access to the original oracle labels as well. \square

4 Full characterisation of LD^f , LDO^f , LD , and LDO

Our goal in this section is to complete the characterisation of the power of scalar oracles with respect to the classes LD and LDO . We aim at giving a robust characterisation that holds also for minor variations in the definition of a scalar oracle. In particular, all of the key results can be adapted to weaker oracles that only receive an upper bound $N \geq n$ on the size of the graph.

4.1 Large oracles can be stronger than identifiers

Let us first consider large oracles. By prior work [13] and Lemma 3 we already know that for any computable large oracle f we have a linear order

$$\text{LDO} \subsetneq \text{LD} \subseteq \text{LDO}^f = \text{LD}^f.$$

Trivially, there is a large computable oracle $f(n) = (1, 2, \dots, n)$ such that

$$\text{LDO} \subsetneq \text{LD} = \text{LDO}^f = \text{LD}^f.$$

We will now show that there is also a large computable oracle f such that

$$\text{LDO} \subsetneq \text{LD} \subsetneq \text{LDO}^f = \text{LD}^f.$$

For a simple proof, we could consider the large oracle $f(n) = (n, n, \dots, n)$. Now the parity language L that consists of graphs with an even number of nodes is clearly in LDO^f but not in LD . However, this separation is not robust with respect to minor changes in the model of scalar oracles. In particular, if the oracle only knows an upper bound on n , we cannot use the parity language to separate LDO^f from LD .

In what follows, we will show that the *upper bound oracle* f that labels all nodes with some upper bound on $N \geq n$ can be used to separate LDO^f from LD . We resort again to computability arguments. The construction that we use in the proof has some elements that we do not need here (in particular, the fragment collection and the parameter r), but it enables us to directly reuse the same tools that we already introduced in the proof of Lemma 2.

Theorem 4. *For the upper bound oracle f there exists a language L such that $L \in \text{LDO}^f \setminus \text{LD}$.*

Proof. The following language is not in LD but is in LDO^f . We will again use a Turing machine construction similar to the one from Lemma 2. For a Turing machine M , let $J(M)$ be T^Δ , the execution table of M augmented with a quad tree as in the proof of Lemma 2. We augment it so that the pivot receives an extra label $\ell \in \{0, 1\}$. Let

$$L = \{J(M, \ell) : \ell \in \{0, 1\}, \text{ and Turing machine } M \text{ outputs } \ell\}.$$

First, observe that L is in LDO^f . Checking the structure of T^Δ and hence $J(M, \ell)$ is known to be in LDO . Since the execution table of M is contained in $J(M, \ell)$, it must halt within $n \leq N$ steps. Finally, since the pivot is guaranteed to receive an $N \geq n$ as its oracle label, it can simulate M for at most N steps and determine whether M halts with output ℓ .

Next, we show that $L \notin \text{LD}$. Suppose otherwise. Fix a local verifier A that decides L , and let r be the running time of A . Consider a node v that is within distance more than r from the pivot. For such a node, algorithm A must always output *yes*—otherwise we could change the input label ℓ so that an answer *no* is incorrect. Thus one of the nodes within distance r from the pivot must be able to tell whether $J(M, \ell)$ is a *no* instance.

Using A , we can now design a sequential algorithm B that solves the undecidable problem of separating the languages L_i from (1). Given a Turing machine M , algorithm B :

- constructs $J(M, 1)$ up to distance $2r$ from the pivot,

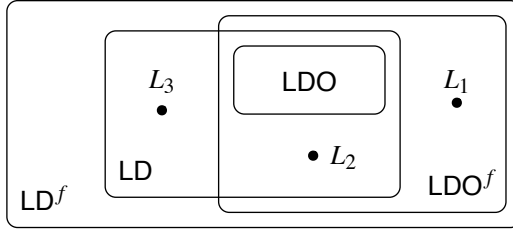


Figure 4: There is a small oracle f such that each of the languages L_i exists.

- assigns the unique identifiers arbitrarily in this neighbourhood,
- simulates A for each node within distance r from the pivot.

Note that B can essentially simulate M for $2r$ steps to construct $J(M, 1)$ up to distance $2r$ from the pivot; the construction is correct if M halts, and it terminates even if M does not halt. Now $J(M, 1)$ is a *no*-instance if and only if M halts with output 0. In this case one of the nodes within distance r from the pivot has to output *no*; otherwise all of them have to output *yes*. In the former case B outputs 0, otherwise it outputs 1. Clearly B outputs ℓ for each $M \in L_\ell$. However, such an algorithm B cannot exist. Therefore A cannot exist, either, and we have $L \notin \text{LD}$. \square

4.2 Small oracles and identifiers are incomparable

In the case of small oracles, we already know that $\text{LDO}^f \subsetneq \text{LD}^f$ for any small oracle f by Lemma 2. Next we characterise the relationship of LDO^f and LD . In essence, we show that these classes are incomparable.

Theorem 5. *There is a single small oracle f so that each of the languages L_1 , L_2 , and L_3 shown in Figure 4 exist.*

Proof. Let f be the small oracle

$$f(n) = (0, 0, \dots, 0, b_n),$$

where b_n is an n -bit string such that the i th bit tells whether the i th Turing machine halts. We construct the languages as follows:

L_1 : Let $P(n)$ denote the labelled path of length n such that each node has two input labels: n and the distance to a specified leaf node v_0 . The correct structure of $P(n)$ is in LDO . Now let

$$L_1 = \{P(M) : \text{Turing machine } M \text{ halts}\}.$$

The node that receives the n -bit oracle label can use it to decide whether the n th Turing machine halts, and therefore $L_1 \in \text{LDO}^f$. Conversely, we have $L_1 \notin \text{LD}$; otherwise we would have a sequential algorithm that solves the halting problem for each Turing machine M by constructing the path $P(M)$ with some fixed identifier assignment and simulating the local verifier.

L_2 : We can use the same language

$$L_2 = \{G(M, r, 0) : r \geq 1 \text{ and Turing machine } M \text{ outputs } 0\},$$

where $G(M, r, 0)$ is defined in the proof of Lemma 2. It is known that $L_2 \in \text{LD}$ and $L_2 \notin \text{LDO}$ [13]. Since checking the structure of $G(M, r, 0)$ is in LDO , it suffices to note that the node that receives the bit vector b_n of length n can use the *length* of the vector as an upper bound in simulating M . Thus $L_2 \in \text{LDO}^f$.

L_3 : Apply Lemma 2. □

We conclude by noting that Theorem 5 is also robust to minor variations in the definitions. In particular, the oracle does not need to know the exact value of n ; it is sufficient that at least one node receives the bit string b_N , where $N \geq n$ is some upper bound on n .

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