ON THE SEQUENTIAL NATURE OF UNIFICATION

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▷ The problem of unification of terms is log-space complete for P. In deriving this lower bound no use is made of the potentially concise representation of terms by directed acyclic graphs. In addition, the problem remains complete even if infinite substitutions are allowed. A consequence of this result is that parallelism cannot significantly improve on the best sequential solutions for unification. However, we show that for the problem of term matching, an important subcase of unification, there is a good parallel algorithm using $O(\log^2 n)$ time and $n^{O(1)}$ processors on a PRAM. For the $O(\log^2 n)$ parallel time upper bound we assume that the terms are represented by directed acyclic graphs; if the longer string representation is used we obtain an $O(\log n)$ parallel time bound. ◁

1. INTRODUCTION

Unification is an important step in resolution theorem proving [16] with applications to a variety of symbolic computation problems. In particular, unification is used in PROLOG interpreters [3], type inference algorithms [12], and term rewriting systems [8]. Many symbol manipulation problems are inherently difficult and thus do not have efficient solutions. Theorem provers and PROLOG interpreters do not always give us the answers we want fast enough. One way to combat the difficulty of these problems is by coordinating many processors to solve a single problem instance by working on several subproblems in parallel. Although there are a number of ways to introduce parallelism into interpreters [17] and theorem provers, unification is a prime target since it is the most commonly repeated operation in these tasks. However, our analysis suggests that parallel unification algorithms will not perform

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significantly faster than the best sequential algorithms known [1], (e.g., [15] runs in linear time). We show that, unless $P \subseteq NC$, an unlikely twist of complexity theory, no parallel algorithm for unification will run in time bounded by a polynomial in the logarithm of the input size, and using a number of processors bounded by a polynomial in the size of the input. We use the PRAM of [5] as our model of parallel computation, although we could, just as well, have used any other "reasonable parallel model" [11].

Informally, two symbolic terms s and t are unifiable if there is some way of substituting additional terms for variables in s and t so that both become the same term. All occurences of a variable x in both s and t must be replaced by the same term. For example, the terms f(x, x) and f(g(y), g(g(z))) may be unified by substituting g(z) for y and g(g(z)) for x. A unification problem like "unify $f(t_1, t_2)$ and $f(t_3, t_4)$" may be decomposed into two subproblems "unify t_1 and t_3" and "unify t_2 and t_4". However, these two problems cannot be solved entirely separately in parallel. If some variable x occurs in both t_1 and t_4, for example, then the solutions to the subproblems must be coordinated so that both substitute the same term for x.

There are several variations of the unification problem. For example, a type inference algorithm may construct labeled graphs which represent terms that must be unified. An acceptable result of unification, in this case, may be a labeled graph with a cycle. Labeled graphs with cycles represent types defined by recursion [14], or, if interpreted as terms, represent "infinite terms". Thus one natural, unrestricted version of unification is to allow "infinite terms" to be substituted for variables. Using the "infinite term" f(f(f...)), we can unify x with f(x), something we could not do otherwise. Unrestricted unification also appears in many PROLOG interpreters; those omitting the *occur* test [3]. Another variation on unification is the special case in which the labeled graphs are from a class of treelike directed acyclic graphs (which we call *simple dags*). The complexity of unification on simple dags is precisely the complexity of unification on symbolic (string) representations of terms, as opposed to the complexity as a function of the size of more concise graph representations. For this case it was known that unification, without "infinite terms", is co-NLOGSPACE-hard [11]. This did not exclude the possibility of parallel algorithms; moreover, no lower bound was known for unrestricted unification.

We show that all of the above variants of unification are log-space complete for P [1, 6, 7, 10], and hence unlikely to have nice parallel solutions. The nondeterministic log-space test for ununifiability in [11], which could have led to a $O(\log^2 n)$ parallel time solution, is sufficient, but unfortunately not necessary [see Figure 3(b) for a counterexample to this test].

One important special case of unification can be solved quickly in parallel. This problem, called term matching, arises in term rewriting. A term s matches a term t if t is a substitution instance of s. The rewrite rule $l \to r$ may be used to rewrite a term t whenever l matches t [8]. We show that matching can be accomplished in $\log^2$-time on a PRAM, using a polynomial number of processors. Our algorithm combines parallel transitive closure of a directed acyclic graph, with parallel computation of connected components of an undirected graph [9, 2]. Also, matching is in NLOG SPACE, and for simple dags it is in DLOGSPACE.

Following the definitions presented in Section 2, we will discuss labeled graph unification in Section 3, unification for simple dags in Section 4, and term matching in Section 5.

2. DEFINITIONS

2.1 Terms and Dags

Let V be an infinite set of *variables* $x, y, z, x_1, \ldots$ and F an infinite set of *function symbols* $f, g, h, f_1, \ldots$ We assume that V and F are disjoint. Each function symbol f has a fixed *arity*, a nonnegative integer a(f). A function symbol $g \in F$ with $a(g) = 0$ is called a *constant*. The set T of *terms* is defined inductively by:

a variable $x \in V$ or constant $g \in F$ is a term, and

if $f \in F$ and $t_1, \ldots, t_{a(f)}$ are terms, then $f(t_1, \ldots, t_{a(f)})$ is a term.

Terms may be represented using directed acyclic graphs with labeled nodes and, possibly, multiple labeled arcs. A *labeled directed graph* is a finite directed graph G, such that:

1. every node v of G has a unique label, denoted label(v), with $\text{label}(v) \in V \cup F$,
2. for each $x \in V$, there is at most one node v with $\text{label}(v) = x$, and it has outdegree 0,
3. if a node v has label $f \in F$, with arity $a(f) \geq 0$, then it has outdegree a(f), and the arcs leaving it are labeled $1, 2, \ldots, a(f)$, respectively.

If there is an arc labeled i from node u to node v, then we say that v is the ith *son* of u. A *labeled dag* G is a labeled directed *acyclic* graph. The *leaves* of G are the nodes of outdegree 0; note that a node v is a leaf iff label(v) is either a variable or a constant. The *height* of a node v of a dag G is the length of the longest path from v to a leaf. A *root* of a dag is a node of indegree 0.

If G is a labeled dag, we can associate a term t_v with any node v of G. We say that v *represents* t_v. The term t_v is defined by induction on the height of v:

if v is a leaf, then $t_v = \text{label}(v)$,

if v has sons $v_1, \ldots, v_k$, and $\text{label}(v) = f$, then $t_v = f(t_{v_1}, \ldots, t_{v_k})$.

The definition of labeled dags above ensures that t_v is always a well-formed term. If G is a labeled directed graph, then we can associate an *infinite term* t_v with each node v of G by a similar definition. Since we only consider finite graphs, all terms represented by nodes of a labeled graph G are finite iff G is acyclic. If G is a labeled dag with only one root r, then we say that G represents the term t_r.

The representation of terms by labeled dags is illustrated in Figure 1. The terms g(x) and x are represented by the two nodes of the labeled dag in Figure 1(a). Both roots in Figures 1(b), 1(c) represent $f(f(x, x), f(x, x))$. The terms $h(x, x, y, z)$ and $h(g(y), g(g(z)), g(g(g_1)), g(g_2))$ are represented by the roots of Figure 1(d). In Figure 1, we assume that $a(f) = 2$, $a(h) = 4$, $a(g) = 1$, and $a(g_1) = a(g_2) = 0$.

Although each node of a labeled dag determines a single term, the converse is not true. A term t can be represented by several different dags. In particular, if t is a term with several occurrences of a subterm t_1, then we may use a separate subdag for each occurrence of t_1 in t, or use one subdag for all occurrences, cf. Figures 1(b) and 1(c). Since a repeated subterm need be represented only once, it is possible to represent some very long terms with relatively small labeled dags. For example, the dag in Figure 1(e) with n nodes represents a term with $O(2^n)$ symbols. We define a class of labeled dags which are no more concise than terms.

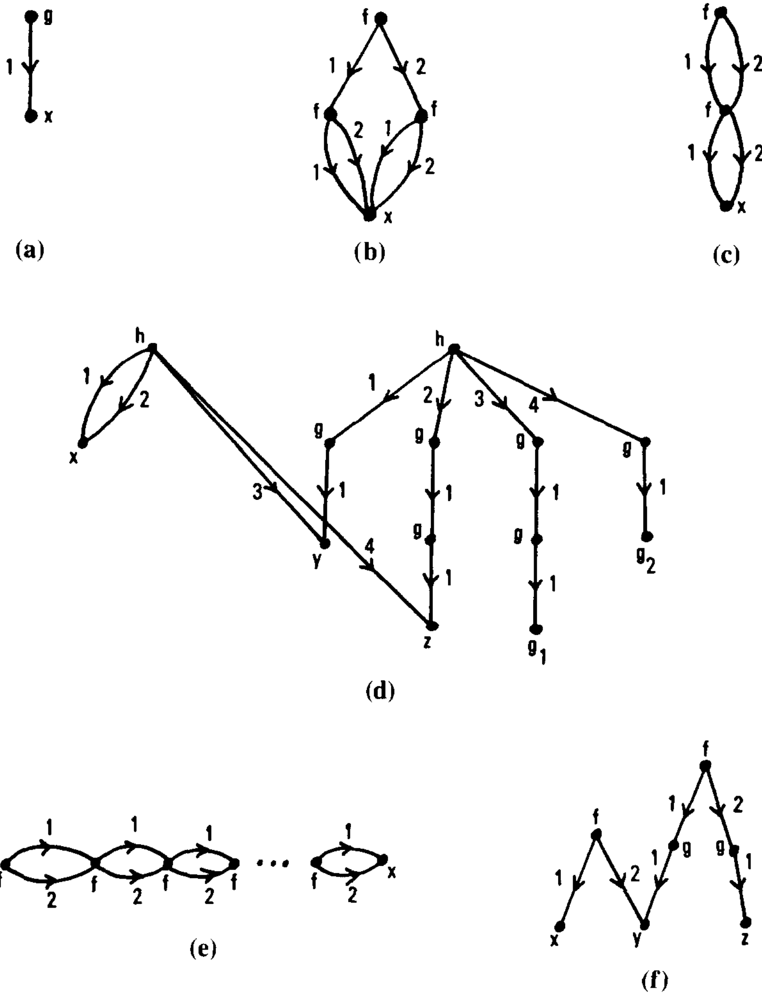


FIGURE 1. Labeled dags.

A *simple dag* is a labeled dag G such that the only nodes of G with indegree greater than 1 are leaves. Thus every node of a simple dag that is not a leaf or a root must have indegree 1. Given a term t (in the form of a string of symbols), we can construct a simple dag representing t in linear time, using only logarithmic space. Similarly, given a simple dag G with a single root, we can write out the term represented by G in linear time and logarithmic space. Moreover, the size of a simple dag, measured in number of nodes and arcs, is within a constant multiplicative factor of the length of the term it represents.

2.2 Unification and Term Matching

Unification and term matching are both problems that are solved by computing substitutions. A *substitution* σ is a mapping from variables to terms such that $\sigma(x) = x$ for all but finitely many $x \in V$. The action of a substitution σ on a term t,

written $\sigma(t)$, is the result of replacing each variable x in t by $\sigma(x)$. Thus $\sigma(f(t_1, \ldots, t_k)) = f(\sigma(t_1), \ldots, \sigma(t_k))$. In particular, any substitution σ maps every function symbol to itself. We use $=$ to denote syntactic equality of strings.

Two terms s and t are *unifiable* if there exists a substitution σ such that $\sigma(s) = \sigma(t)$. A term s *matches* term t if there exists a substitution σ with $\sigma(s) = t$.

In some instances we may wish to allow substitutions to map variables to infinite terms. If we allow these more general substitutions, then we have the *unrestricted unification* and *unrestricted matching* problems. Unrestricted unification differs from unification [e.g., in Figure 1(a) x and g(x) are ununifiable but unrestricted unifiable with $\sigma(x) = g(g(\ldots))$, an infinite term]. Unrestricted matching and matching are the same; note that we only consider substitutions that involve infinite terms, not unification of infinite terms s and t.

If $\sigma(s) = \sigma(t)$, then σ is called a *unifier* for s and t. A substitution σ is *more general* than a substitution τ if there exists a substitution ρ with $\tau = \rho \circ \sigma$. In [16], it is shown that whenever terms s and t are unifiable, there is a unifier σ for s and t, which is more general than any other unifier. This is called the *most general unifier* (mgu) for s and t. The mgu is unique up to renaming of variables. For example, consider the terms $s = f(x, y)$ and $t = f(g(y), g(z))$ represented in Figure 1(f). These terms are unifiable, with mgu $\sigma(x) = g(g(z))$, $\sigma(y) = g(z)$, and $\sigma(z) = z$; then $\sigma(s) = \sigma(t) = f(g(g(z)), g(z))$.

Two terms s and t are unifiable if a certain kind of relation, can be constructed on the nodes of a labeled dag representing s and t. If u and v are two nodes of a labeled dag and if u_i is the ith son of u and v_i the ith son of v, for some i, then u_i, v_i are *corresponding* sons of u, v. A relation R on the nodes of a labeled dag is a *correspondence* relation if, for all u, v, u_i, v_i:

if uRv then $u_i R v_i$ whenever u_i, v_i are corresponding sons of u, v.

A correspondence relation that is also an equivalence relation will be called a *c-e relation*. A relation R is *homogeneous* if label(u) and label(v) whenever uRv are not different function symbols. An equivalence relation R on nodes of a labeled directed graph G is *acyclic* if the R-equivalence classes are partially ordered by the arcs of G. In [15], acyclic, homogeneous c-e relations are called valid equivalence relations. These relations characterize unifiability.

Proposition 1. [14] *Let* u *and* v *be nodes in a labeled dag* G. *Then* t_u *and* t_v *are unrestricted unifiable iff there is a homogeneous* c-e *relation* R, *with* uRv. *Similarly,* t_u *and* t_v *are unifiable iff there is an acyclic, homogeneous* c-e *relation* R, *with* uRv.

If R is an acyclic, homogeneous c-e relation on a labeled dag G, then the reduced graph formed by treating each equivalence class as a single node is again a labeled dag. If u and v are the only two roots of G, and uRv, then this reduced graph with a single root represents a term s that is a substitution instance of both t_u and t_v. If R is the *minimal* c-e relation with uRv, then $s = \sigma(t_u) = \sigma(t_v)$, where σ is the mgu of t_u and t_v [15]. We can extract σ from R by taking $\sigma(x)$ to be the term in the reduced graph that is represented by the node formed from the equivalence class of x. We can therefore consider the reduced labeled dag as a reasonable representation of a unifier for two terms. This representation of a unifier has the virtue of being

compact: it is clear that the reduced graph is no larger than the original dag. However, if we were to write out each unifier explicitly, we might end up writing out terms that were much longer than the terms represented by the input dag. An example in [15] shows that the length of the substitution may be an exponential function of the length of the input terms.

As in [15], we will represent equivalence relations on the nodes of labeled dags by adding undirected edges to the labeled dag data-structure.

Matching may be viewed as a special case of unification. Let σ_c be a substitution such that for each distinct variable x, in the terms we are examining, $\sigma_c(x)$ is c_x, a distinct constant symbol not appearing in these terms. It is easy to see that a term s matches a term t iff s and $\sigma_c(t)$ are unifiable. Another degenerate case of unification is to determine whether two terms are syntactically identical. Of course, this is a trivial operation on strings, but it is not quite so trivial an operation when terms are represented by labeled dags. Clearly, s and t are syntactically equal iff $\sigma_c(s)$ and $\sigma_c(t)$ are unifiable.

In summary, using the labeled dag data structure, we have the following problems:

UNIFY(G, u, v)

Input: A labeled dag G with distinguished nodes u and v.

Output: Are t_u and t_v unifiable? If yes, then produce a labeled dag representing the mgu.

MATCH(G, u, v): This is UNIFY(G, u, v) with $\sigma_c(t_v)$ instead of t_v.

EQUAL(G, u, v): This is UNIFY(G, u, v) with $\sigma_c(t_u)$, $\sigma_c(t_v)$ instead of t_u, t_v.

Of course, there is also unrestricted unification $\text{UNIFY}^\infty(G, u, v)$. We have a special case of each of the above problems when G is a simple dag.

2.3 Parallelism, NC and P

For sequential computation we use the standard definitions for time, space, space-bounded reductions, and complexity classes such as P, DLOGSPACE, NLOGSPACE, T(n)-DSPACE, on a Random Access Machine (RAM) [1]. We denote log-space reducibility by $\leq_{\log}$. As usual, P is the class of languages recognizable in deterministic polynomial time. The problems UNIFY, MATCH, and EQUAL all belong to P [13, 15]. Some may be solved in $\log^{O(1)} n$ space, while others, those log-space complete for P, most probably cannot.

For parallel computation we use the Parallel RAM (PRAM) of [5] as our model, with parallel time and number of processors as the critical resources. We make use of the *parallel computation thesis*, relating parallel time and sequential space, and its proof for PRAM's [5]:

$$\bigcup_{k \geq 0} \log^k(n)\text{-parallel time-PRAM} = \bigcup_{k \geq 0} \log^k(n)\text{-DSPACE}.$$

We take NC to be the class of problems solvable on a PRAM using $\log^{O(1)} n$ parallel time, and $n^{O(1)}$ processors. We try to determine whether a problem in P is "parallelizable" (i.e., in NC) or "most probably not parallelizable" (i.e., log-space complete for P); [1] reviews related results.

One problem that is log-space complete for P is the circuit value problem for monotone circuits. A *monotone circuit* β is a sequence $(\beta_0, \ldots, \beta_n)$, where each β_i is either an input, an AND gate AND(j, k), or an OR gate OR(j, k); where for indices j, k we have $i > j > k$, and the 0, 1 values of the inputs are given explicitly. In addition, monotone circuits are assumed to have the following properties:

1. if β_i is an input, then the index i appears at most once in β, (fan-out ≤ 1 for inputs),
2. if β_i is a gate, then the index i appears at most twice in β, (fan-out ≤ 2 for gates),
3. β_n is an or-gate with one output.

The *monotone circuit value problem* is:

MCV = $\{\beta | \beta$ is a monotone circuit with the output value of $\beta_n = 0\}$.

From [6, 7] we have:

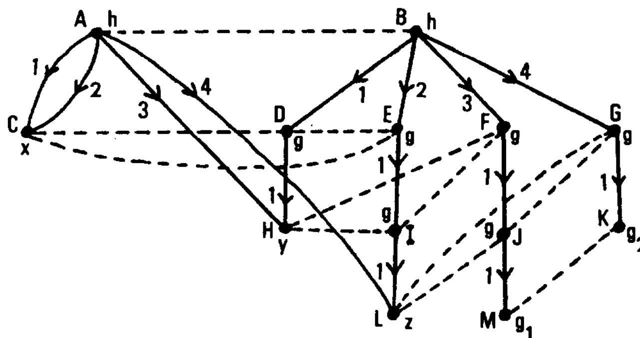
Proposition 2. MCV *is log-space complete for* P.

3. THE COMPLEXITY OF UNIFICATION

The general unification problem, encountered in theorem proving and elsewhere, is to find a unifier for a set of terms. However, the general case is log-space and linear time reducible to the special case of unifying a single pair of terms [15]. On a PRAM this reduction can be performed in O(log n) parallel time and with O(n) processors; it affects none of our results.

We first describe a naive unification algorithm based on the criterion of Proposition 1, and on the fact that the mgu is the minimal c-e relation [15]. The input to the algorithm is a labeled dag G with two distinguished nodes u and v. We wish to solve UNIFY(G, u, v). A relation ® is constructed and maintained as undirected edges in G. The relation ® is by its representation symmetric and reflexive. In order to make ® a c-e relation, both "correspondence" and "equivalence" must be satisfied. Setting

FIGURE 2. Illustrating naive-unification. (1) A®B. (2) C®D, C®E, H®F, L®G (propagation). (3) D®E (transitivity). (4) H®I (propagation). (5) F®I (transitivity). (6) L®J (propagation). (7) J®G (transitivity). (8) M®K (propagation). (9) Ununifiable because M and K have distinct labels g_1 and g_2.



sons equivalent, when their fathers are equivalent, is known as *propagation*. For ® to be an equivalence relation we must also enforce *transitivity*. Having created the minimal c-e relation ® for which u®v, we then test for homogeneity. In the affirmative case a new labeled graph G′ can be constructed by coalescing classes of nodes in G. Now we know that the input is at least unrestricted unifiable. If G′ is acyclic it is unifiable.

proc naive-unification(G, u, v)

set u®v;

while (® is not a c-e relation) **do**

propagation: **while** (u®v have corresponding sons u_i, v_i not related by ®) **do**
  set u_i®v_i **od**;

transitivity: **while** (u®v and v®w, but u, w are not related by ®) **do**
  set u®w **od od**;

if ® not homogeneous **then** print UNUNIFIABLE

  else {coalesce equivalence classes to produce labeled graph G′}

   if G′ has a cycle

   then print UNUNIFIABLE BUT UNRESTRICTED UNIFIABLE

   else print UNIFIABLE

   fi

fi

corp {G′ represents mgu}

In this algorithm all individual steps can be performed on a PRAM using $\log^{O(1)} n$ time and $n^{O(1)}$ processors. The difficulty arises in the outer loop, the body of which is executed if ® is either not a correspondence, or not an equivalence relation, i.e., if either condition inside an inner loop is satisfied. The problem is that on an input of size n the body of the main loop might be executed $\Omega(n)$ times. This behavior is illustrated in Figure 2. The example can easily be generalized to force the $\Omega(n)$ alternation between propagation and transitivity for any n.

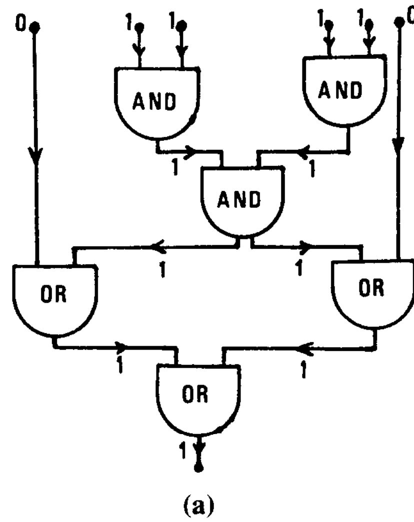
Theorem 1. UNIFY(G, u, v) *and* UNIFY$^{\infty}$(G, u, v) *are log-space complete for* P.

PROOF. We show how to log-space reduce MCV to unifiability (for membership see [15]). More specifically, if α is a monotone circuit $\{\alpha_0, \alpha_1, \dots, \alpha_n\}$, we construct G($\alpha$), u($\alpha$), and v($\alpha$) such that

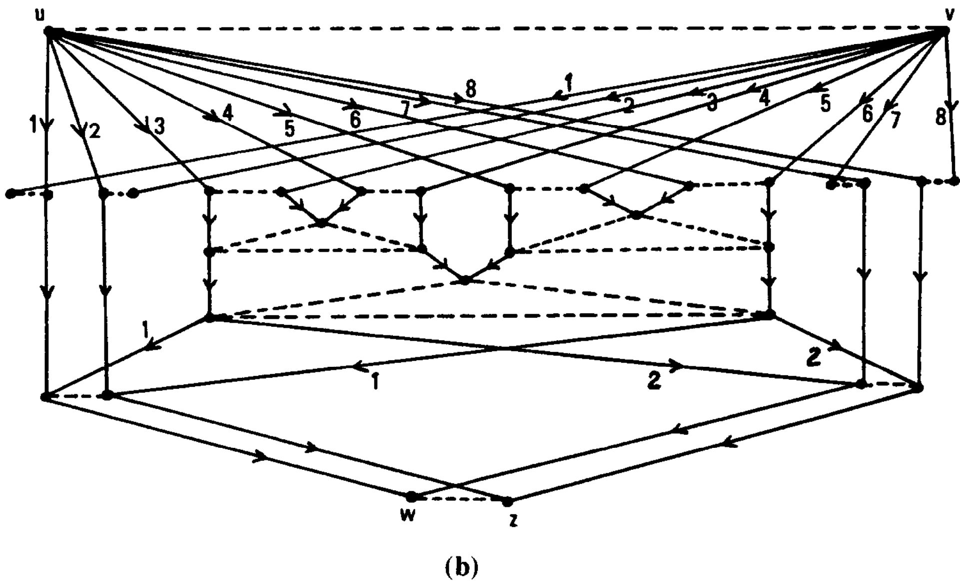
$$\alpha \in \text{MCV iff UNIFY}(G(\alpha), u(\alpha), v(\alpha)) = \text{UNIFIABLE}.$$

This reduction also applies to UNIFY$^{\infty}$ and is easily seen to use only log space.

The monotone circuit α can be represented as a diagram with wires, AND and OR gates of fan-in 2 and fan-out at most 2, a special OR output gate with one output wire, and with each input wire leading to one gate and having a 0 or a 1 value [see Figures 3(a) and 3(b) for an example]. The input wire values combine to produce



(a)

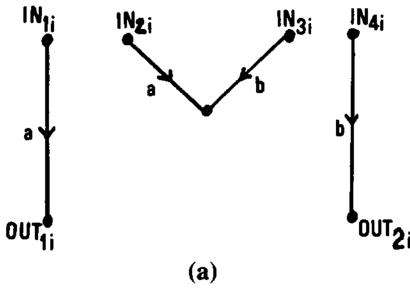


(b)

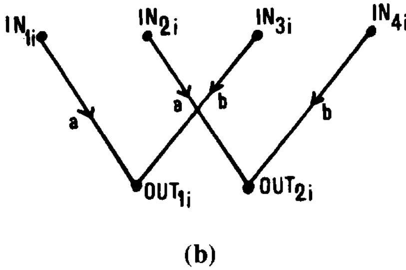
FIGURE 3. Label(u) = label(v) = h, label(w) = $g_1 \neq g_2$ = label(z); label(node of outdegree 1) = g; label(node of outdegree 2) = f; label(leaf other than w, z) = distinct variable.

values for all other wires and the output wire in particular. The circuit has no feedback, i.e., if the wires are viewed as arcs and the inputs and gates as nodes we get a dag without multiple arcs.

1. Introduce two nodes $u(\alpha)$, $v(\alpha)$ in $G(\alpha)$.
2. If α_i is an AND gate include G_{and} from Figure 4(a) in $G(\alpha)$. If α_i is an OR gate include G_{or} from Figure 4(b) in $G(\alpha)$. These dags have two pairs of input nodes and one pair of output nodes each, i.e., $\{IN_{1i}, IN_{2i}\}$, $\{IN_{3i}, IN_{4i}\}$, and



(a)



(b)

FIGURE 4. Theorem 1 subgraphs. (a) G_{and}. (b) G_{or}.

$\{OUT_{1i}, OUT_{2i}\}$. Corresponding sons are illustrated by the labels a, b on the arcs.

3. If α_i is an input include in $G(\alpha)$ a pair of nodes $\{OUT_{1i}, OUT_{2i}\}$. If the value of the input is 1 then make OUT_{1i}, OUT_{2i} corresponding sons of $u(\alpha), v(\alpha)$. If the value of the input is 0 then make OUT_{1i}, OUT_{2i} sons of $u(\alpha)$ and let $v(\alpha)$ have two sons that correspond to them and are two new leaves in $G(\alpha)$.
4. If gate α_i is connected to α_j, α_k (i.e., in the wire diagram) then identify nodes $IN_{1i} = OUT_{1j}, IN_{2i} = OUT_{2j}, IN_{3i} = OUT_{1k}, IN_{4i} = OUT_{2k}$. When these subdags are concatenated, nodes have outdegree ≤ 2, and the labels on the arcs can be made 1 and 2, so that the equalities of labels a, b in Figures 4(a) and 4(b) are preserved.
5. In the dag constructed in Steps 1–4 above assign labels to the nodes as follows:

$$\text{label}(u) = \text{label}(v) = h,$$

$$\text{label}(\text{node of outdegree } 1) = g,$$

$$\text{label}(\text{node of outdegree } 2) = f,$$

$$\text{label}(OUT_{1n}) = g_1 \neq g_2 = \text{label}(OUT_{2n}),$$

$$\text{label}(\text{leaf other than } OUT_{1n}, OUT_{2n}) = \text{distinct variable}.$$

We can easily see now that every wire w in the wire diagram can be associated to a pair of nodes $OUT_w, OUT_{w'}$. We require $u(\alpha)$®$v(\alpha)$. For such a minimal c-e relation ®, we claim that the value of wire w in α is 1 iff OUT_w®$OUT_{w'}$. This certainly holds for the inputs, because of the way we built corresponding sons of u and v. Also, it is trivial to check that G_{and} and G_{or} simulate the behavior of AND and OR gates. Therefore the value of α_n is 1 iff OUT_{1n}®OUT_{2n}. The graph $G(\alpha)$ is

constructed in such a way that the only place homogeneity could be violated by ® is if OUT_{1n}®OUT_{2n}. As a result, if $\alpha = 1$, the terms represented by $u(\alpha)$ and $v(\alpha)$ are not unrestricted unifiable, and if $\alpha = 0$ they are unifiable (the acyclicity condition is also true). □

4. SIMPLE DAGS

In this section we will make our lower bounds independent of the potentially concise dag representation of terms, by extending them to simple dags.

Theorem 2. UNIFY(G, u, v) *and* UNIFY^{∞}(G, u, v) *are log-space complete for* P, *even when* G *is a simple dag.*

PROOF. Given monotone circuit α we construct a simple dag $G(\alpha)$ with two roots $u(\alpha)$, and $v(\alpha)$ so that, if $\alpha_n = 0$ then the terms $t_{u(\alpha)}, t_{v(\alpha)}$ are unifiable or else they are not unrestricted unifiable. This suffices for the completeness of both UNIFY and UNIFY^{∞}. Note that the proof of Theorem 1 no longer applies, because the G_{or} dags used in that reduction could introduce nodes with indegree 2, i.e., their output nodes, which were not leaves.

As in the proof of Theorem 1, we encode the input of α using a pair of nodes for each circuit input. The input-subgraph of the graph of Theorem 1 is actually a simple dag, so we use the same construction. However, we cannot attach "gates" directly to the input-subgraph since this will produce a dag which is not simple. Instead, each gate will be constructed separately using a pair of subgraphs. Any c-e relation ® with $u(\alpha)$®$v(\alpha)$ will relate the two parts of each gate. In addition, the input nodes of one gate will be "connected" to input-subgraph nodes or output nodes of other gates using a separate "patch board" subgraph. Recall that the gates of α are numbered so that if an output of gate α_i goes to an input of gate α_j, then $i < j$.

For each gate of α, we use four input nodes and four output nodes. For gate α_i, let us denote these nodes by $IN_{1i}, \ldots, IN_{4i}$ and $OUT_{1i}, \ldots, OUT_{4i}$. As in the proof of Theorem 1, the nodes of $G(\alpha)$ work in pairs. Inputs IN_{1i} and IN_{2i} represent the first input to α_i and IN_{3i} and IN_{4i} the second. Similarly, nodes OUT_{1i} and OUT_{2i} represent the first output of α_i and OUT_{3i} and OUT_{4i} the second. We also use nodes u_i, v_i which are the ith sons of roots $u(\alpha)$ and $v(\alpha)$, respectively, and four or seven internal nodes which may remain anonymous.

If α_i is an OR gate, then we construct a simple dag $GATE_i$ as in Figure 5(a), with u_i, v_i corresponding sons of $u(\alpha)$, $v(\alpha)$. If ® is a c-e relation with $u(\alpha)$®$v(\alpha)$, it is easy to see that OUT_{1i}®OUT_{2i} and OUT_{3i}®OUT_{4i} if either IN_{1i}®IN_{2i} or IN_{3i}®IN_{4i}. It will be clear from the construction of $G(\alpha)$ that if ® is minimal, then these are the only cases in which the output nodes will be related by ®. If α_i is an AND gate, similar reasoning applies for the simple dag of Figure 5(b) which simulates the logic of AND.

The remaining task is to "connect" the gates so that if, for example, the first output of α_i goes to the second input of α_j, then IN_{3j}®IN_{4j} whenever OUT_{1i}®OUT_{2i}. We use an example connection between α_i and α_j to illustrate the construction of a "patch board" simple dag PATCH, which contains two new nodes u_p, v_p and IN and OUT nodes from the input-subgraph and gate subgraphs of $G(\alpha)$. Let u_p, v_p be corresponding sons of $u(\alpha), v(\alpha)$, different from the sons used in the gate and input

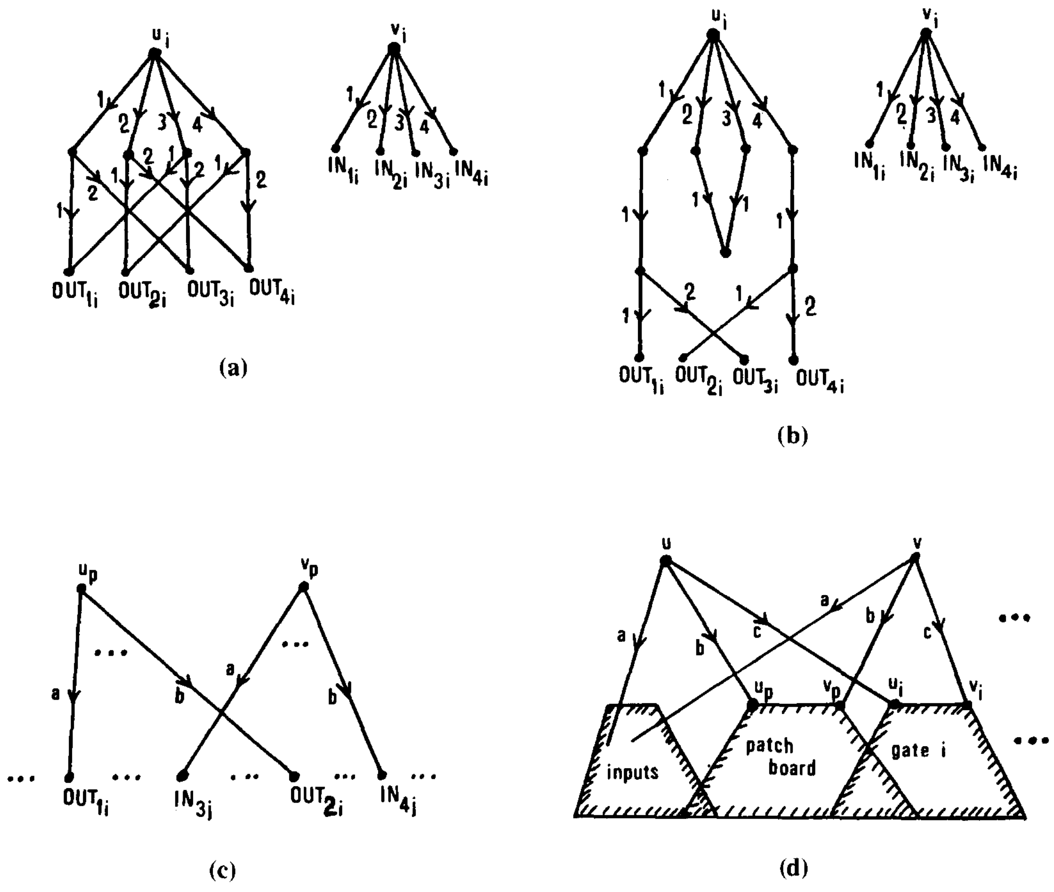


FIGURE 5. Theorem 2 subgraphs. (**a**) OR subgraph. (**b**) AND subgraph. (**c**) Example use of PATCH. (**d**) Putting everything together.

subgraphs. Now make IN_{3j} and OUT_{1i} corresponding sons of u_p and v_p; also make IN_{4j} and OUT_{2i} corresponding sons of u_p and v_p [see Figures 5(c) and 5(d)]. When $u(\alpha) \circledR v(\alpha)$, two input nodes of $GATE_j$ will be merged if the right two output nodes of $GATE_i$ are.

As in the proof of Theorem 1, we label the outputs of the final gate with different constant symbols. All other nodes have labels that depend on their arity, so that nodes with outdegree 2, say, have the same label. It is easy to verify by induction that in the minimal c-e relation $\circledR$ with $u(\alpha) \circledR v(\alpha)$, we have $OUT_{1n} \circledR OUT_{2n}$ and $OUT_{3n} \circledR OUT_{4n}$ iff the output of the last gate α_n is 1. This completes the proof of Theorem 2. □

5. A PARALLEL ALGORITHM FOR TERM MATCHING

Unification is a practical sequential algorithm for matching since unification can be done in linear time. However, unification is not a good parallel approach to matching. We show how MATCH(G, u, v) can be computed in $\log^2 n$ parallel time

using polynomial many processors. In addition, we prove that MATCH(G, u, v) is in co-NLOGSPACE. If G is a simple dag then MATCH(G, u, v) is actually in DLOGSPACE.

When we wish to determine whether s matches t, we will assume w.l.o.g. that no variables appear in t. In Section 6 we further clarify the relationship between matching and unification. Since MATCH(G, u, v) is the same as UNIFY(G, u, v) when no variables appear in t_v, we know that t_u matches t_v iff there is a homogeneous c-e relation $\sim$ on G with $u \sim v$. A refinement of this characterization of term matching suggests an efficient parallel algorithm.

Lemma 1. Let G *be a labeled dag with nodes* u *and* v, *and let the subgraph of* G *induced by the descendants of* v *have no nodes labeled with variables. Let* R *be the minimal correspondence relation on* G *with* uRv, S *be the minimal equivalence relation containing* R, *and* T *be the minimal correspondence relation containing* S. *Then* t_u *matches* t_v *iff* T *is homogeneous.*

PROOF. If t_u matches t_v then since t_u and t_v are unifiable, the minimal c-e relation $\sim$ with $u \sim v$ is homogeneous. Since $\sim$ must contain T, it follows that T is homogeneous.

For the converse, suppose that T is homogeneous. We will define a substitution σ such that $\sigma(t_u) = t_v$. Let G_u, G_v be the subgraphs of descendants of u, v, respectively. We first show that for every node x in G_u there is a node y in G_v such that xRy. If, on the contrary, there is some x in G_u without xRy for any y in G_v, then let w be the last node in some path from u to x with wRz for some z in G_v. Since w has a son, label(w) is a k-ary function symbol for some $k > 0$. By similar reasoning, label(z) is a zero-ary function symbol. But then label(w) $\neq$ label(z) and hence T is not homogeneous. It follows from this contradiction that every S-equivalence class contains at least one node from G_v.

For each S-equivalence class E, pick some node e in E from G_v. If w is another G_v node in E, then since T is homogeneous and no variables appear in G_v, we can argue that $t_w = t_e$ (here we have the problem EQUAL). We now define the substitution σ. For any variable x in t_u, let E be the S-equivalence class of the node labeled x and define $\sigma(x) = t_e$. It is easy to check by induction on the height of a node w in G_u that if wRz, then $\sigma(t_w) = t_z$. Thus $\sigma(t_u) = t_v$ and t_u matches t_v. □

Given any relation, we can find the minimal correspondence relation R containing it, in $\log^2 n$ parallel time and $n^{O(1)}$ processors on a PRAM, using a transitive closure algorithm [2]. If G is a labeled dag with n nodes, we define an n^2 by n^2 boolean *correspondence matrix* C_G. We associate each (unordered) pair of nodes of G with a row and a column of C_G and define the entries of C_G:

$$C_G(\{u,v\}, \{x,y\}) = 1 \quad \text{iff} \quad \text{x and y are u and v or corresponding sons of u and v.}$$

Lemma 2. Let G *be a labeled dag with nodes* u *and* v, *and let* R *be the min. correspondence relation s.t.* uRv. *Then* xRy *iff the* ({u, v}, {x, y}) *entry of* C_G*'s transitive closure equals* 1.

Now given relation R, we can find the minimal equivalence relation S containing R using a connected components algorithm on the rows of C_G. It is well known that

connected components can be computed in $\log^2 n$ parallel time and $n^{O(1)}$ processors on a PRAM [2, 17].

Since computing correspondence relations twice, connected components once, and testing for homogeneity are sufficient to decide matching, we have that MATCH(G, u, v) can be computed in $\log^2 n$ parallel time and $n^{O(1)}$ processors on a PRAM (or equivalently MATCH $\in$ NC).

In fact, we can show somewhat tighter complexity upper bounds, since DLOGSPACE $\subseteq$ NLOGSPACE $\subseteq$ NC:

Theorem 3. The set of $\langle G, u, v\rangle$ *such that* MATCH(G, u, v) = *false is in* NLOGSPACE. *Furthermore, if* G *is a simple dag, then this recognition problem is in* DLOGSPACE.

PROOF. Let G be a dag with MATCH(G, u, v) = *false*. Let R, S, T be relations on the nodes of G as in the statement of Lemma 1. By Lemma 1, there must be nodes x and y of G such that xTy, but label(x) and label(y) are two different function symbols. We show that there is a log-space bounded nondeterministic Turing machine M_T capable of guessing all pairs (x, y) such that xTy, and checking whether x and y have the same labels. Thus, recognizing the $\langle G, u, v\rangle$, such that, MATCH(G, u, v) = *true* is a problem in co-NLOGSPACE (also a subset of the class NC).

To begin with, let M_R be a nondeterministic machine that starts with the pair (u, v) on its worktape. A move of M_R consists of replacing a pair (x, y) with a pair (x_i, y_i) of corresponding sons of x and y. Clearly M_R is capable of guessing (x, y) iff xRy.

We now define a nondeterministic machine M_S using M_R. The machine M_S begins by running M_R some nondeterministic number of steps to guess a pair (x, y). Subsequently, M_S repeats the following 3 steps nondeterministically:

1. If one pair (x, y) or two pairs (x, y)(w, z) are on the worktape, then it may replace (x, y) by (y, x).
2. If (x, y), (y, z) are on the worktape, then it may replace both by the single pair (x, z).
3. If the single pair (x, y) is on the worktape, then it may run M_R some number of steps to guess (w, z) and end up with both pairs (w, z), (x, y) on the worktape.

With these primitive steps M_S may guess (x, y) iff xSy.

Finally, we build M_T from M_S. This machine behaves just like M_R, but instead of starting with (u, v), starts with any pair (x, y) that M_S is capable of guessing. This concludes the proof of the first part of the theorem, which describes the PRAM algorithm sketched above, from the point of view of nondeterministic log-space.

If G is a simple dag, then M_R can easily be made a deterministic depth-first enumerator of pairs (x, y). This machine MD_R always maintains the pair immediately preceding the current one, so that it can backtrack from leaf nodes. Backtracking from internal nodes is straightforward since each has indegree 1.

Using a log-space preprocessor we can treat the subgraph rooted at v as a tree. Recall that this graph has no variables, so that all we need to do is duplicate leaves labeled with constants. By doing this we limit the number of times Step 2 of M_S must be repeated to only two. Thus we can construct a deterministic machine MD_S that enumerates all (x, y) such that xSy. Finally, we build a deterministic MD_T from MD_R and MD_S as before. □

A corollary of Theorem 4 is that for simple dags deciding whether MATCH(G, u, v) = *true* is also in DLOGSPACE, since DLOGSPACE is closed under complement. From the analysis in [5] it also follows that this problem can be solved in O(log n) parallel time on a PRAM.

6. CONCLUSIONS AND OPEN PROBLEMS

We have demonstrated that several versions of unification are complete for P. This suggests, by way of the parallel computation thesis, that unification is inherently sequential. It is unlikely that significant improvements in the speed of theorem provers, interpreters for logic programs, and the like will be brought about by the development of parallel unification algorithms. However, for the special case of term matching, the prospects are much brighter. Term matching can be accomplished in log n or $\log^2 n$ parallel time, depending on whether the input is in the form of a simple dag.

We might also point out that unification of terms s and t is complete for P even if s and t do not contain any variables in common (this is different from t having no variables). Also, if s and t are unifiable this does not imply that s matches t or that t matches s. However, if s matches t then s and t are unrestricted unifiable. If s matches t and t matches s they are unifiable.

The problem of computing the *congruence closure* of a relation [4, 10] appears to be a directional dual of the unification problem. In [10] computing the congruence closure of a relation is shown to be log-space complete for *P*. The complexity of other algebraic word problems, which may be viewed as generalizations of the unification problem, is also examined in [10].

In the sequential case congruence closure seems slightly harder than unification [4]. Perhaps because there are remarkable similarities between the sequential algorithms for unification and testing equivalence of deterministic finite automata. However, the *inequivalence* of deterministic finite automata can be detected nondeterministically using only logarithmic space. A machine can see that two automata A_1 and A_2 are inequivalent by guessing an input string, character by character, and simulating the actions of both machines as it goes. If one ends up in an accept state while the other rejects, then the two are clearly different. If A_1 and A_2 differ, then some sequence of characters must surely uncover this. Thus unification is subtly, but fundamentally different from this "almost identical" problem.

In [18] H. Yasuura, using different techniques, has independently derived Theorem 1 of Section 3.

Some interesting open problems remain unresolved, namely: (1) lower bounds for the complexity of MATCH and EQUAL, or can our upper bounds be improved, (2) the number of processors used in the transitive closure of a correspondence matrix is unrealistically large, and it would be of some practical significance to decrease it to even n^3, and finally (3) what is the complexity of *commutative* matching, i.e., if function symbols stand for commutative operations.

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