بسم الله الرّحمن الرّحيم

دانشگاه صنعتی اصفهان _ دانشکدهٔ مهندسی برق و کامپیوتر (نیمسال تحصیلی ۴۰۲۲)

طراحي الگوريتمها

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unit propagation

The real value of simplification is that it enables a further optimization that can drastically decrease the search space.

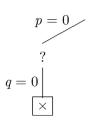
In the course of the backtracking search, if we see a sentence that consists of single atom, say p, we know that the only possible satisfying assignments further down the branch must set p to true. In this case, we can fix p to be true and ignore the subbranch that sets p to false. Similarly, when we encounter a sentence that consists of a single negated atom, say $\neg p$, we can fix p to be false and ignore the other subbranch. This optimization is called unit propagation because sentences of the form p or $\neg p$ are called units.

To start, let p be false:



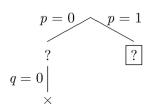
Original	Simplified
$p \lor q$	q
$p \vee \neg q$	$\neg q$
$\neg p \lor q$	_
$\neg p \vee \neg q \vee \neg r$	_
$\neg p \lor r$	_

In the simplified set of sentences, we have the unit $\neg q$, so we fix q to be false (unit propagation). (We also have the unit q, so we could have fixed q to true. The result is the same in either case.)



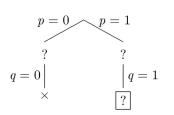
Original	Simplified
$p \vee q$	false
$p \vee \neg q$	_
$\neg p \lor q$	_
$\neg p \vee \neg q \vee \neg r$	_
$\neg p \lor r$	_

 Δ is falsified, so we backtrack to the most recent decision point, all the way back at the root. Let p be true.



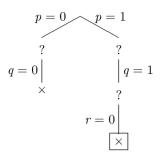
Original	Simplified
$p \vee q$	-
$p \vee \neg q$	_
$\neg p \lor q$	q
$\neg p \vee \neg q \vee \neg r$	$\neg q \vee \neg r$
$\neg p \lor r$	r

In the simplified set of sentences, we have the unit q so we do unit propagation, fixing q to be true. (We could also have performed unit propagation using the other unit r.)



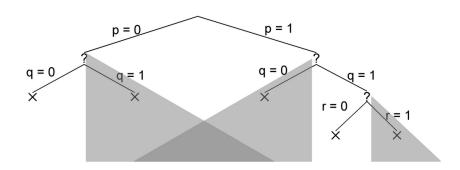
Original	Simplified
$p \vee q$	_
$p \vee \neg q$	-
$\neg p \lor q$	_
$\neg p \vee \neg q \vee \neg r$	$\neg r$
$\neg p \lor r$	r

In the simplified set of sentences, we have the unit $\neg r$ so we do unit propagation, fixing r to be false.



Original	Simplified
$p \vee q$	_
$p \vee \neg q$	_
$\neg p \lor q$	_
$\neg p \vee \neg q \vee \neg r$	_
$\neg p \lor r$	false

All branches are closed, so the method determines that Δ is unsatisfiable. Compared to the tree explored by the basic backtracking search, we see that the greyed out subtrees are pruned away from the search space.



Pure symbol heuristic

A pure symbol is a symbol that always appears with the same "sign" in all clauses. For example, in the three clauses

$$(A \vee \neg B) \wedge (\neg B \vee \neg C) \wedge (C \vee A)$$

the symbol A is pure because only the positive literal appears, B is pure because only the negative literal appears, and C is impure. It is easy to see that if a sentence has a model, then it has a model with the pure symbols assigned so as to make their literals true, because doing so can never make a clause false. Note that, in determining the purity of a symbol, the algorithm can ignore clauses that are already known to be true in the model constructed so far. For example, if the model contains B=0, then the clause $(\neg B \lor \neg C)$ is already true, and in the remaining clauses C appears only as a positive literal; therefore C becomes pure.

DPLL

The Davis-Putnam-Logemann-Loveland method (DPLL) is a classic method for SAT solving. It is essentially backtracking search along with unit propagation and pure literal elimination. Most modern, complete SAT solvers are based on DPLL, with additional optimizations not discussed here. These SAT solvers are routinely used to solve SAT problems with large numbers of propositions.

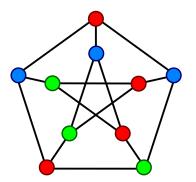
```
proc DPLL (F: clause set): bool
// outputs 1, if F is satisfiable, 0 otherwise
if \Box \in F then return 0
if F = \emptyset then return 1
if F contains a unit clause \{u\} then return DPLL(F\{u=1\})
if F contains a pure literal u then return DPLL(F\{u=1\})
choose with the adequate strategy a variable x \in Var(F)
if DPLL (F\{x=0\}) then return 1
return DPLL (F\{x=1\})
```

Graph Coloring

توجه کنید که این برای دومین بار است که مسئلهٔ رنگ آمیزی گراف مواجه می شویم. بار اول (در فصل راهبرد حریصانه) با صورت بهینهسازی آن مواجهه داشتیم: یعنی یافتن کمترین تعداد از رنگها که به وسیلهٔ آنها می توان رئوس یک گراف را رنگ کرد. اما این بار با صورت تصمیم گیری آن روبرو هستیم: آیا می توان رئوس یک گراف را تنها با استفاده از حداکثر m رنگ، رنگ آمیزی کرد یا خیر m یک عدد صحیح داده شده و معین است. این صورت از مسئله را مسئله m رنگ آمیزی گراف می نامیم.

مسئلهٔ ۲_رنگ آمیزی گراف یک مسئلهٔ سخت نیست، و برای حل نمونه های این مسئله، الگوریتمی با مرتبهٔ چندجمله ای نسبت به تعداد رئوس گراف موجود است. اما اگر $m \geq 3$ ، آنگاه مسئلهٔ mرنگ آمیزی گراف در حالت عمومی NP-complete

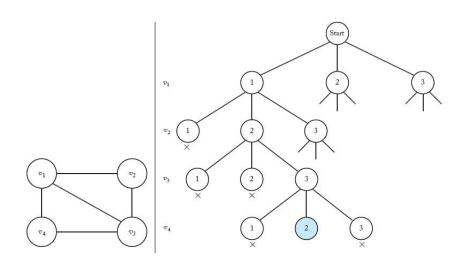
The m-Coloring problem concerns finding all ways to color an undirected graph using at most m different colors, so that no two adjacent vertices are the same color. We usually call the m-Coloring problem a unique problem for each value of m.



State Space Tree

A straightforward state space tree for the m-Coloring problem is one in which each possible color is tried for vertex v_1 at level 1, each possible color is tried for vertex v_2 at level 2, and so on until each possible color has been tried for vertex v_n at level n. Each path from the root to a leaf is a candidate solution. We check whether a candidate solution is a solution by determining whether any two adjacent vertices are the same color. To avoid confusion, remember in the following discussion that "node" refers to a node in the state space tree and "vertex" refers to a vertex in the graph being colored.

We can backtrack in this problem because a node is nonpromising if a vertex that is adjacent to the vertex being colored at the node has already been colored the color that is being used at the node. The number in a node is the number of the color used on the vertex being colored at the node. The first solution is found at the shaded node. Nonpromising nodes are labeled with crosses.



After v_1 is colored color 1, choosing color 1 for v_2 is nonpromising because v_1 is adjacent to v_2 . Similarly, after v_1 , v_2 , and v_3 have been colored colors 1, 2, and 3, respectively, choosing color 1 for v_4 is nonpromising because v_1 is adjacent to v_4 .

The Backtracking Algorithm for the m-Coloring Problem

ightharpoonupProblem: Determine all ways in which the vertices in an undirected graph can be colored, using only m colors, so that adjacent vertices are not the same color.

Inputs: positive integers n and m, and an undirected graph containing n vertices. The graph is represented by a two-dimensional array W, which has both its rows and columns indexed from 1 to n, where W[i][j] is true if there is an edge between ith vertex and the jth vertex and false otherwise.

ors, so that no two adjacent vertices are the same color. The output for each coloring is an array vcolor indexed from 1 to n, where vcolor[i] is the color (an integer between 1 and m) assigned to the ith vertex.

```
void m_coloring (index i)
  int color:
  if (promising(i))
    if (i == n)
       cout << vcolor[1] through vcolor[n];
     else
        for (color = 1; color \le m; color++){ // Try every
           vcolor[i+1] = color;
                                             // color for
           m_{-}coloring(i + 1);
                                                 // next vertex.
bool promising (index i)
 index j;
  bool switch;
  switch = true;
  j = 1;
                                            // Check if an
  while (i < i \&\& switch){
     if (W[i][j] \&\& vcolor[i] == vcolor[j]) // adjacent vertex
      switch = false;
                                              // is already
                                               // this color.
     j++;
  return switch:
```

Following our usual convention, n, m, W, and vcolor are not inputs to either routine. The top level call to $m_coloring$ would be $m_coloring(0)$.

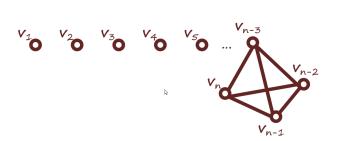
The number of nodes in the state space tree for this algorithm is equal to

$$1 + m + m^2 + m^3 + \dots + m^n = \frac{m^{n+1} - 1}{m - 1}$$

For a given m and n, it is possible to create an instance that checks at least an exponentially large number of nodes (in terms of n).

مثال (عملكرد ناكارآمد)

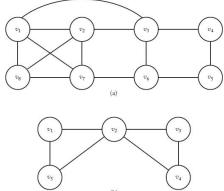
الگوریتم ما برای آنکه بفهمد گراف زیر \mathbf{n}_{-} رنگ پذیر نیست، درخت فضای حالتی میسازد که تعدادی نمایی نود دارد (نسبت به n). چرا؟



تمرین: یک گراف همبند با n رأس بسازید (برای هر n داده شدهٔ دلخواه) که الگوریتم ما برای پیبردن به این حقیقت که آن گراف \mathbf{r}_- رنگ پذیر نیست، مجبور به ملاقات تعدادی نمایی نود باشد.

The Hamiltonian Circuits Problem

Given a connected, undirected graph, a Hamiltonian Circuit (also called a tour) is a path that starts at a given vertex, visits each vertex in the graph exactly once, and ends at the starting vertex.



A state space tree for this problem is as follows. Put the starting vertex at level 0 in the tree; call it the zeroth vertex on the path. At level 1, consider each vertex other than the starting vertex as the first vertex after the starting one. At level 2, consider each of these same vertices as the second vertex, and so on. Finally, at level n-1, consider each of these same vertices as the (n-1)st vertex.

The following considerations enable us to backtrack in this state space tree:

- 1. The ith vertex on the path must be adjacent to the (i-1)st vertex on the path.
- 2. The (n-1)st vertex must be adjacent to the 0th vertex (the starting one).
- 3. The *i*th vertex cannot be one of the first i-1 vertices.

The Backtracking Algorithm for the Hamiltonian Circuits Problem

- Problem: Determine all Hamiltonian Circuits in a connected, undirected graph.
- Inputs: positive integer n and an undirected graph containing n vertices. The graph is represented by a two-dimensional array W, which has both its rows and columns indexed from 1 to n, where W[i][j] is true if there is an edge between the ith vertex and the jth vertex and false otherwise.
- Outputs: For all paths that start at a given vertex, visit each vertex in the graph exactly once, and end up at the starting vertex. The output for each path is an array of indices vindex indexed from 0 to n-1, where vindex[i] is the index of the ith vertex on the path. The index of the starting vertex is vindex[0].



```
bool promising (index i)
 index i:
 bool switch;
 if (i == n-1 \&\& !W[vindex[n-1]] [vindex[0]])
    switch = false;
                                    // First vertex must be adjacent
 else if (i > 0 \&\& !W[vindex[i-1]]
    switch = false; [vindex[i]])
                                      // to last. ith vertex must
 else {
                                      // be adjacent to (i-1)st.
    switch = true;
    j = 1;
    while (j < i \&\& switch){ // Check if vertex is
        if (vindex[i] == vindex[j]) // already selected.
           switch = false;
        i++:
  return switch;
```

Following our convention, n, W, and vindex are not inputs to either routine. If these variables were defined globally, the top-level called to hamiltonian would be as follows:

$$vindex[0] = 1;$$
 (ست، v_1 است) ما v_1 است) $hamiltonian(0);$

The number of nodes in the state space tree for this algorithm is

$$1 + (n-1) + (n-1)^{2} + (n-1)^{3} + \dots + (n-1)^{n-1} = \frac{(n-1)^{n} - 1}{n-2},$$

which is much worse than exponential.

مثال (عملكرد ناكارآمد)

Although the following instance does not check the entire state space tree, it does check a worse-than exponential number of nodes. Let the only edge to v_1 be one from v_2 , and let all the vertices other than v_1 have edges to each other. There is no Hamiltonian Circuit for the graph, and the algorithm will check a worse-than-exponential number of nodes to learn this.