

Depth-First Search and Its Use Case in Distributed Systems Debugging

Nate Kremer-Herman



Depth-first search kernel

Given a large graph. Start at a root node. Find all reachable vertices. Measured in TEPS, just like BFS.

Worst case performance: O(|V| + |E|) 17

8

9

2

6

3

5

4

Iterative pseudocode

- 1 **procedure** DFS-iterative(*G*,*v*):
- 2 let *S* be a stack
- 3 S.push(v)

9

- 4 while S is not empty
- 5 *v* = *S*.pop()
- 6 **if** *v* is not labeled as discovered:
- 7 label v as discovered
- 8 **for all** edges from v to w **in** G.adjacentEdges(v) **do**
 - S.push(w)

Implementation techniques

- Implemented in Perl
 - Regex matching
 - Data structures
- Used the iterative algorithm
 - Can only use recursion to a certain depth
 - Faster albeit less elegant
- Is essentially a bottom-up DFS (kinda)
 - I cheat and make the leaves the roots



```
For Perl
wizards:
```

```
sub iterative
   my (\$v) = @;
   my $return = "";
   my @stack;
   push(@stack, $v);
   while(scalar(@stack)) {
       my $n = pop(@stack);
       if(($nodes{$n}{'v'} != $i)) {
           \frac{1}{v'} = i;
           $return = $return . "$n:";
           $traversed++;
           my @children = split(":", $nodes{$n}{'c'});
           my @attrs = split(":", $nodes{$n}{'a'});
           foreach my $c (@children) {
               my @cattrs = split(":", $nodes{$c}{'a'});
               my cf = 0;
               foreach my $ca (@cattrs) {
                   foreach my $na (@attrs) {
                       if(substr($na, 1) eq substr($ca, 1))
                           if (\$c \ge 0) { push (@stack, $c);
                           last;
                   if($cf) { last; }
   return $return;
```



Notional summary (for everyone else)

- 1 **procedure** DFS-iterative(*G*,*v*):
- 2 push v on a stack, S
- 4 while S is not empty
- 5 *v* = S.pop()
- 6 **if** *v* has not been visited in this round:
- 7 label *v* as visited
- 8 for all child edges of v do
- 9 **if** *child* has a matching attribute with *v*:
- 10 S.push(child)

Updated complexity analysis

- Algorithm is still O(|V| + |E|)
 - Worst case, we look at all vertices
 - Best case, we look at no vertices
- Added overhead for attribute analysis
 - We only keep vertices which share attributes (for debugging)
 - Attribute checking slows down traversal by an order of magnitude or two (fun)

Metrics

- Measured performance in TEPS
- Captured error nodes in separate graphs
 - Examples to come

Datasets

- All datasets are currently synthetic
 - Each is a binary graph
 - Generated via Perl script
- Number of nodes ranges from 10 1,000,000
 - Realistic dataset size O(100) O(10,000)
 - Tiny: 10 nodes
 - Small: 100 nodes
 - \triangleright

. . .

- Colossal: 1,000,000 nodes
- Any bigger runs into memory limitations







Much more manageable output per failed task.





Sometimes a task fails on its own, not because of a previous task.

N16 [A=2]

N34 [:A=1]



13

Implementation performance results

Only Traversal		Traversal + Computation	
Nodes	TEPS	Nodes	TEPS
10	361,347.81	6 (10)	60,676.46
100	375,014.11	71 (100)	74,231.57
1,000	367,753.35	872 (1,000)	69,404.00
10,000	500,512.47	9,724 (10,000)	118,423.95
100,000	476,622.42	99,032 (100,000)	99,258.17
1,000,000	458,047.21	998,270 (1,000,000)	120,060.52

Plans for parallel implementation

- Use Work Queue master-worker framework to parallelize traversal
 - Cascading traversal pattern
 - May not be faster than serial implementation for a realistic dataset (resource acquisition)
- Use real data if there is time
 - Only roadblock is transforming debug logs into graphs, traverser is done
 - Each type of log has its own syntax, so each requires a handwritten parser



Questions?

