Project of Algorithms on Node Labeling

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1 Introduction

In this project we discuss the "Node Labeling Problem", in which we attempt to label the nodes of a graph with unique labels from a set of labels. We have the following definitions:

- Let G = (V, E) be an undirected graph.
- Let d(u, v) be the distance between nodes u and v.
- For all nodes $v \in V$, let $N(v, h) \subseteq V$ be the set of nodes that are at most h hops away from v.
- Let $K = \{0, 1, \dots, k-1\}$ be the set of k integers, where $k \leq |V|$.
- For all $v \in V$, let $c(v) \in K$ be the label of node v, where different nodes may have the same label.
- Let C(v,h) be the set of labels of nodes in N(v,h).
- A labeling of the nodes is valid if every label in K is used at least once.
- Let r(v) be the smallest integer such that the node v has all the labels in K in N(v, r(v)).
- Let m(v) be the smallest integer such that the node v has at least k nodes in N(v, m(v)).

Formally, our relevant sets and values can be defined as follows:

$$N(v,h) \triangleq \{u \in V \mid d(u,v) \le h\}$$

$$C(v,h) \triangleq \{c(u) \mid u \in N(v,h)\}$$

$$r(v) \triangleq \min\{h \mid |C(v,h)| = k\}$$

$$m(v) \triangleq \min\{h \mid |N(v,h)| \ge k\}.$$

Note that in general, we have $|C(v,h)| \leq |N(v,h)|$, since the labels of nodes in N(v,h) are not necessarily distinct, and $r(v) \geq m(v)$, since there must be at least one node per label but not necessarily one label per node.

The Node-Labeling Decision Problem is defined as follows:

Given:

- An undirected graph G = (V, E)
- A set of $k \le |V|$ labels $K = \{0, 1, ..., k 1\}$
- A nonnegative integer R,

does there exist a labeling c(v) for all $v \in V$ such that |C(v,R)| = k for all $v \in V$?

Now consider this as an optimization problem. The Node-Labeling Optimization Problem is defined as follows:

Given:

- An undirected graph G = (V, E)
- A set of $k \le |V|$ labels $K = \{0, 1, \dots, k-1\}$,

find a valid labeling for all the nodes such that $\max_{v \in V} \frac{r(v)}{m(v)}$ is minimized.

In the case of the optimization problem, if an algorithm that solves it has $\max_{v \in V} \frac{r(v)}{m(v)} \leq \rho$ for all possible instances, then we say that the algorithm has a proximity ratio of ρ , and the algorithm is a ρ -proximity algorithm.

First, we will prove that the Node-Labeling Decision Problem is NP-Complete. Then, we will present a polynomial-time algorithm for the Node-Labeling Optimization Problem where the graph is a tree, analyze the proximity ratio of the algorithm, and finally analyze the runtime complexity of the algorithm.

2 NP-Completeness Proof

Theorem 1. The Node-Labeling Decision Problem is NP-Complete.

Proof. A problem is NP-Complete if it is in NP and every problem in NP can be reduced to it in polynomial time. We will show the former by presenting a polynomial time algorithm to verify a solution to the Node-Labeling Decision Problem, and the latter by reducing k-coloring to the Node-Labeling Decision Problem.

First, consider the following algorithm to verify a solution to the Node-Labeling Decision Problem:

```
Algorithm 1: Verify a Solution to the Node-Labeling Decision Problem
```

Algorithm 2: BFS

```
Input: A node v, current depth d, set of labels seen l
Output: True if all k labels are seen within depth R, False otherwise if d > R then

| return False
end
| l = l \cup \{c(v)\} if |l| = k then
| return True
end
for each neighbor u of v do
| if BFS(u, d+1, l) then
| return True
| end
end
return False
```

This algorithm works by performing a breadth-first search from each node v in the graph, and checking if all k labels are seen within depth R. If a depth of R is reached without seeing all k labels, the algorithm returns False. Otherwise, the algorithm returns True. The algorithm runs in $O(|V| \cdot (|V| + |E|))$ time, which is polynomial in the size of the input. Thus, the Node-Labeling Decision Problem is in NP.

Now we perform a reduction from the 3-coloring problem to the Node-Labeling Decision Problem. Let G' = (V', E') be an instance of the 3-coloring problem. We can construct an instance of the node labeling problem (G = (V, E), K, R) as follows:

- V = V'
- \bullet E = E'
- $K = \{0, 1, 2\}$

This transformation can be done in polynomial time, since we only copy the graph and set K and R to constant values. We claim that G' is 3-colorable if and only if G has a valid labeling such that |C(v,R)| = 3 for all $v \in V$.

- (\Rightarrow) Assume that G' is 3-colorable. Then there exists some valid 3-coloring $c':V'\to\{0,1,2\}$ of G'. Let $c:V\to K$ be the labeling of G such that c(v)=c'(v) for all $v\in V'$. Since c' is a valid 3-coloring, it uses all the colors, so c will also be a valid labeling of G. Then, for all $v\in V$, N(v,2) must contain at least 3 nodes, so in order for the labeling to be valid, at least 3 distinct labels must be used. Thus, |C(v,2)|=3 for all $v\in V$.
- (\Leftarrow) Assume that there exists a valid labeling $c:V\to K$ of G such that |C(v,2)|=3 for all $v\in V$. We can use the same coloring c' of G' such that c'(v)=c(v) for all $v\in V'$. For all edges $(u,v)\in E'$, we have that $u\in N(v,2)$ and $v\in N(u,2)$. Since |C(u,2)|=3 and |C(v,2)|=3, we have that $c'(u)\neq c'(v)$, or else one of them would only see 2 distinct labels in 2 hops. Thus, c' is a valid 3-coloring of G'.

Now we have shown that the Node-Labeling Decision Problem is in NP and that the 3-coloring problem can be reduced to it in polynomial time. Therefore, the Node-Labeling Decision Problem is also NP-hard, and thus NP-Complete.

3 Approximation Algorithm

Here we discuss an algorithm to solve the Node-Labeling *Optimization* Problem when the input graph is a tree.

4 Pseudocode

```
Input: Adjacency list of a tree adj\_list, number of labels k
Output: A labeling of the nodes
LabelNodes (adj\_list, k, max\_attempts) best\_labeling \leftarrow null;
best\_max\_ratio \leftarrow \infty;
for i \leftarrow 1 to max\_attempts do
       initial\_labels \leftarrow InitialLabeling(adj\_list, k);
        final\_labeling \leftarrow \texttt{ImproveLabeling}(adj\_list, k, initial\_labels);
       ratios \leftarrow \texttt{CalculateProximityRatios}(adj\_list, k, final\_labeling);
       max\_ratio \leftarrow max(ratios);
       if max\_ratio < best\_max\_ratio then
                best\_labeling \leftarrow final\_labeling;
                best\_max\_ratio \leftarrow max\_ratio;
       end
       if max\_ratio = 1.0 then
         return final_labeling;
       end
end
return best_labeling;
\texttt{InitialLabeling}(\textit{adj\_list}, \textit{k}) \textit{ } n \leftarrow |\textit{adj\_list}|;
labeling \leftarrow [-1, \ldots, -1];
                                                                                                                                      // Length n
for i \leftarrow 0 to n-1 do
       if labeling[i] = -1 then
               available\_labels \leftarrow \{0, \dots, k-1\} \setminus \{labeling[j] : j \in
                  adj\_list[i], labeling[j] \neq -1\};
                labeling[i] \leftarrow random choice from available\_labels (or from
                   \{0,\ldots,k-1\} if available\_labels is empty);
       \quad \text{end} \quad
end
return labeling;
ImproveLabeling(adj\_list, k, labeling) n \leftarrow |adj\_list|;
for i \leftarrow 1 to n \cdot k do
       ratios \leftarrow \texttt{CalculateProximityRatios}(adj\_list, k, labeling);
       max\_ratio \leftarrow max(ratios);
       if max\_ratio = 1.0 then
               break;
       end
       worst\_nodes \leftarrow \{v : ratios[v] = max\_ratio\};
       worst\_node \leftarrow random choice from worst\_nodes;
        distances \leftarrow BFS(adj\_list, worst\_node);
       if max\_ratio = \infty then
               neighborhood \leftarrow \text{keys of } distances;
        else
              neighborhood \leftarrow \{node : distances[node] \leq [max\_ratio]\};
       end
       label\_counts \leftarrow count occurrences of each label in neighborhood;
        most\_common\_label \leftarrow_l label\_coants[l];
       least\_common\_label \leftarrow_l label\_counts[l];
       nodes\_with\_most\_common \leftarrow \{v \in neighborhood : labeling[v] = about the common of the c
          most\_common\_label\};
        node\_to\_swap \leftarrow_{v \in nodes\_with\_most\_common} ratios[v];
       labeling[node\_to\_swap] \leftarrow least\_common\_label;
return labeling;
BFS(adj\_list, start) distances \leftarrow \{start : 0\};
queue \leftarrow [(start, 0)];
while queue is not empty do
       node, dist \leftarrow \text{dequeue from } queue;
```

- 5 Proximity Ratio Analysis
- 6 Runtime Complexity Analysis