A Maximum Weighted Flow Algorithm Based on the Push-Relabel Method
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INTRODUCTION

Abstract: Network flows studies how to move entities or objects, such as electrical currents, goods, vehicles, and network packets, from one point in a network to another point efficiently, while utilizing the underlying network resources cost-effectively. Coupling deep theoretical rigor and remarkable range of applicability, network flows spans over several broad disciplines including operations research, computer science, and engineering. In particular, the classic maximum flow problem finds the maximum amount of flow that can be sent from source node to sink node in a flow network, considering that edges in the flow network have capacities that restrain amount of flows on each edge. It has wide real world applications such as baseball elimination, airline scheduling, job scheduling, and network routing. The push-relabel algorithm is one of the most efficient algorithms for the maximum flow problem. Push operation and relabel operation are two basic operations used in the algorithm. In particular, during the execution of the algorithm, it maintains a "preflow" and gradually converts it into a maximum flow by moving flow locally between neighboring vertices using push operations under the guidance of an admissible network maintained by relabel operations.

MOTIVATION

This research was inspired by Dr. Andrew Goldberg’s work in graph theory and optimization. He created the Push-Relabel Algorithm which aims to maximize the amount of entities or objects that can be moved from a source node to a terminal node.

For example, scientists must continuously monitor an active volcano and its environment to accurately assess its conditions. Scientific instruments are installed near the volcano and collect as much data as they can. Although many different types of data are collected, not all data is vital to know if a volcano is about to erupt. Due to energy depletion and storage constraints, not all data will be preserved. Consequently, scientists must somehow filter out unnecessary data and make sure the most vital data is preserved. In order to preserve the most vital data scientists must prioritize the data (flows) and give it a priority value (weight values). For example, data collected near a volcano may include: seismic data, temperature, infrasonic data, and wind speed. Different priority values will be given to each data: 4, seismic data; 3, temperature; 2, infrasonic data; 1, wind speed.

OBJECTIVES

- Analyze the Push-Relabel Algorithm which aims to maximize the amount of entities or objects that can be moved from the source node to a terminal node.
- Implement the new Maximum Weighted Flow Algorithm which aims to maximize the entities or objects with biggest priorities.

RESULTS

Generated a weighted edge and weight graph with these properties:
- Total Number of Nodes = 10
- Total Number of Edges = 14
- Number of different flows = 3
- Priority Value of Node 7 = 3
- Priority Value of Node 8 = 5
- Priority Value of Node 9 = 4

Figure 3 uses a standard algorithm to find the maximum flow which produces a total flow value of 54. In Figure 4 we used the Maximum Weighted Flow Algorithm that takes into consideration flows with priorities and produces the total value of 64 which is larger than the previous algorithm.

METHODS

Hi-Level Push-Relabel (HPHR) Program (Figure 2)

Algorithm: Push-Relabel Method
0. Notations:
   \( e(u): \) node u’s excess flow
   \( h(u): \) node u’s height
   \( c(u, v): \) residual capacity of (u, v)
   \( \beta(u, v): \) residual capacity of (u, v)
1. if \( e(u) > 0 \)
2. while \( e(u) > 0 \) there exists \( (u, v) \) s.t.
3. \( h(u) = h(u) + 1 \) and \( c(u, v) > 0 \)
4. \( y = \min(e(u), c(u, v)) \) through
5. \( (u, v) \) by sending a message to \( v \)
6. \( e(u) = e(u) - y \), \( c(u, v) = c(u, v) - y \)
7. update \( c(u, v) \)
8. end while;
9. \( \beta(u, v) > 0 \)
10. \( h(u) = 1 + \min(h(w), \beta(u, w)) > 0 \)
11. Broadcast \( h(u) \) to neighboring nodes
12. end if;
13. end if;
14. RETURN

Algorithm: Maximum Weighted Flow
0. Notations:
   \( e(u): \) node u’s excess flow
   \( h(u): \) node u’s height
   \( c(u, v): \) residual capacity of (u, v)
   \( \beta(u, v): \) node u’s priority

Super source outputs weighted flows
\( \mathbb{G} = (\mathbb{G}_0, \mathbb{G}_1), \) where \( \mathbb{G}_0 = \{1, 2, 3, 4, 5, 6, 7, 8, 9, 10\}

Sort flows:
\( h(u) = \sum_{i=1}^{10} \beta(u, i) \)
Input flows into directed graph based on highest priority

CONCLUSIONS

- Currently in the process of implementing our ideas
- Finalize the program
- Compare to unweighted flow Push-Relabel algorithm with priority value based algorithm to test performance
- Java Implementation

REFERENCES

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