A Graph Search implementation using C++ threads and FastFlow

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1 Introduction

The purpose of this report is to analyze the speedup which can be achieved using the parallelization of the BFS (Breadth-First Traversal) search traversal given the randomly generated Acyclic Directed Graph (G=(V, E)). Given the value of vertex $s \in V$ which serves as the starting point (source node) of BFS search, the aim of the project is to count the number of times a specific vertex $v \in V$ was found. The BFS search traversal will include the following solutions:

- 1. Sequential BFS traversal
- 2. BFS traversal using standard C++ threads
- 3. BFS traversal using FastFlow

In order to generate a graph (G = (V, E)), a Erdős-Rényi model was used which has been implemented in a small Python script. The ultimate goal is to compare the performances of all above-mentioned solutions and analyze their efficiency.

2 Solution

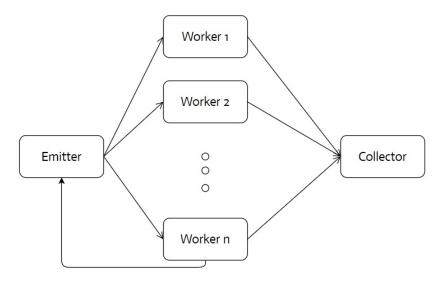
2.1 Sequential BFS traversal

The algorithm of BFS traversal based on the notion of an adjacency list is implemented using following steps:

- 1. the stl queue (queue < int > q) is initialized before traversal begins and initial vertex is pushed into the queue
- 2. within while loop the front element is popped off q.pop()
- 3. if starting vertex $s \in V$ is poped from queue, then all vertices connected to vertex $s \in V$ in this pass are accessed
- 4. if the vertex has not yet been visited **visited**[n] != 1, the processing goes further on
- 5. the vertex is marked as visisted visited[n] = 1
- 6. if vertex **n** equals to $v \in V$, the counter is increased by one
- 7. the current vertex \mathbf{n} is pushed to the queue $\mathbf{q.push}(\mathbf{n})$
- 8. new iteration of loop is started

2.2 C++ thread

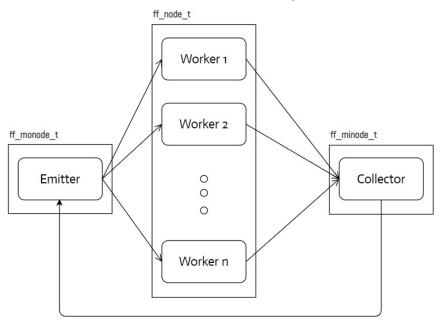
The following algorithm is based on the above-mentioned algorithm of traversing a graph. However, to efficiently increase the parallelization technique *Farm* is used in the context of this solution. There are three main elements used in this solution: *Emitter*, *Worker* and *Collector*.



The load balancing is implemented by using round-robin scheduling which means that emitter (queue<int> q) sends one element of data to each worker at a time. Within each worker, the vertex is checked if it has been visited before (visited[n] != 1). If it hasn't, it is marked as visited (visited[n] = 1). This particular element is pushed to the queue srcToFarm.push(n) and returned to the Emitter to keep on traversing the graph. Once the traversal is over, EOS value is sent to all Workers to finish the computation. The Collector also gets EOS value and keeps on increasing countEos variable to the value equal of the number of workers. Once it is reached, the computation is over.

2.3 FastFlow

The FastFlow solution is designed in the very same way as the one implemented in C++ thread. However, there are slight differences which are important to mention and reflected in the design.



As it is shown in the picture above, there are 3 key elements which are added to the solution: monode, node, minode. It means that solution consists of three blocks. monode matches the role of Emitter and distributes the elements to the Worker. node matches the role of Worker and rettrieves all child elements. minode matches the role of Collector gathers all child elements and forwards them back to Emitter

3 Results

In this section of the report, the results of the solution are discovered to provide a clear picture of the observation. The number of nodes n of the graph G = (V, E) are [10.000, 50.000, 100.000]. The results are generated by runing each solution 5 times to get an average value of the BFS time traversal. The parallelization degree is 2^w where $w \in [0, 1, 2, 3, 4, 5, 6, 7, 8]$. The graph density according to Erdős-Rényi model is p = 0.5 The experiment are conducted on the Xeon PHI machine. Within this report, following metrics

are used to retrieve a necessary information: **speedup**, **scalability** and **efficiency**.

$$speedup(w) = \frac{Tseq}{Tpar(w)}$$

$$\sigma(w) = \frac{Tpar(1)}{Tpar(w)}$$

$$\varepsilon(w) = \frac{Tseq(1)}{w*Tpar(w)}$$

3.1 10k nodes

The starting node $s \in V$ is 0. The node $v \in V$ which is searched is 17. The results of experiment are shown in Table 1

Parallelism degree	Sequential BFS (μ s)	C++ Thread (μs)	FastFlow (μs)
1	133	231	152
2	134	127	88
4	134	60	55
8	134	43	38
16	133	50	33
32	133	50	56
64	134	110	95
128	133	130	224
256	134	195	869

Table 1: BFS traversal on 10k nodes

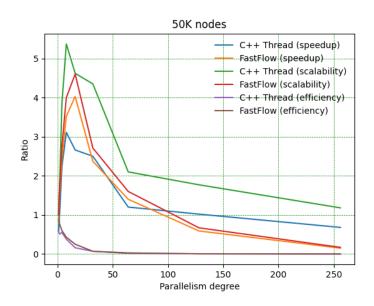


Figure 1: Metrics based on C++ thread and FastFlow

3.2 50k nodes

Parallelism degree	Sequential BFS (μ s)	C++ Thread (μs)	FastFlow (μs)
1	1002	1620	1061
2	1003	871	625
4	1003	480	514
8	1003	261	515
16	1002	263	215
32	1003	225	215
64	1003	201	205
128	1002	361	336
256	1003	427	762

Table 2: BFS traversal on 50k nodes

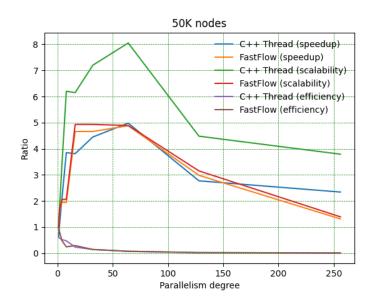


Figure 2: Metrics based on C++ thread and FastFlow

3.3 100k nodes

Parallelism degree	Sequential BFS (μs)	C++ Thread (μs)	FastFlow (μs)
1	2258	3601	2382
2	2258	1955	1382
4	2260	1040	1150
8	2259	747	1136
16	2258	489	1120
32	2261	382	680
64	2259	359	390
128	2260	466	740
256	2260	646	1271

Table 3: BFS traversal on 100k nodes

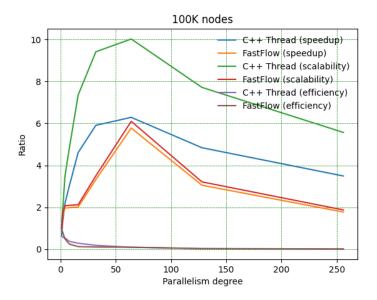


Figure 3: Metrics based on C++ thread and FastFlow

4 Overhead Analysis

As it can be observed from Tables 1, 2 and 3 above and Figures 1, 2 and 3, **speedup** and **scalability** drop as the **parallelism degree** increases. There are multiplicity of reasons:

- Due to the fact that the *Emitter* implements static round-robin scheduling policy, the traversal might take a longer time to complete in case of the particular node having many children while other other threads have to stay idle. This, in turn, contributes to longer **Completion Time**
- In both C++ threads and FastFlow implementations, synchronization policies used as **unique_lock** and **lock_guard** allow to accept one element at a time. As the number of threads grow exponentially, the coordination among threads using synchronization also increases exponentially. Which, in turn, impacts overall **Completion Time**
- As each thread alone without any code implementation and synchronization require a time for its initialization, the increasing number of threads will increase the overall time of execution. As it has been tested without any code implementation and synchronization, on average, each thread require 7 milliseconds. For instance, the code with

parallelism degree of 256 requires approximately 36 milliseconds to be initialized.

• Both solution implement **vectorization**. It means as the **parallelism degree** grows exponentialy, the **overhead** grows alongside. As it is shown on figures 1, 2, 3, the overall *speedup* and *efficiency* fall dramatically with higher number of **parallelism degree**.

5 Instructions

In the root folder **Makefile** is setup for compiling. Instructions for compiling:

- make GraphFF compile FastFlow implementation
- make Graph compile C++ thread implementation

Instructions for running:

- $\bullet \ ./executable_name \ [text_file] \ [start_node] \ [search_node] \ [workers_num]$
- \bullet executable_name graph_search_ff or graph_search
- $text_file$ the file names are: random_10000.txt, random_50000.txt, random_100000.txt
- start_node the node value from which BFS search starts
- search_node the node value which is searched by BFS
- workers_num the number of workers