# Concurrency-Aware Tree/Graph Traversal Algorithms 

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## Background

- Parallelism in data storage
- Multiple drive: Distributed, RAID
- Single drive: Solid-State Disk (SSD)


Exploiting the internal parallelism of SSDs

- SSD has internal concurrency
- Can handle multiple requests simultaneously
- Serial operations cannot achieve best performance
- Parallel algorithm for data
- Tree/Graph structure widely used in data and file system
- Traversal \& Searching Algorithms


## Algorithm - Serial BFS

Breadth-first Search

Layer by layer


Implement by a queue (FIFO)


Pop_front() and push_back() its all descendants

## Parallel BFS

Nodes in one layer can be processed in parallel
Straightforward
$\square$

Need to distinguish current layer and next layer

Frontier and Next Queue


Loop frontier in parallel, push descendants in next

Frontier $=$ Next

## Serial DFS

Depth-first Search
$\square$

Use Stack (FILO)


Pop_back() and push_back()
More common in searching

## Parallel DFS

Each processing relies on last result
Both operation are on stack top, have to wait
A strict order DFS cannot be paralleled efficiently


PDFS:
Unordered/Pseudo Parallel DFS: DFS for each thread
Not globally depth-first, but still prefer depth


## Parallel DFS - cont.

How to distribute nodes to threads?
Keep all threads busy for most parallelism.

- Set a threshold size (fSize) for stack
- Each thread works on own stack
- When one thread's stack is larger than threshold
- Split into two part
- Given one part to a new thread


## Thread 1

Thread 2

Thread 3

We let the OpenMP automatically handle the scheduling. For better performance, control scheduling and balancing.

## Code

## OpenMP: Compiler-level directives, no/minimum

 changes on codes.
## Serial and Parallel BFS:

```
std::vector<int> frontier, next;
bool isFound = false
while (!isFound && frontier.size() > 0):
    for offset in frontier:
        node = read_node(offset)
        if (node.key == key) isFound = true;
        if (isFound || node's children count == 0)
            continue
        next.insert(node.children);
    frontier = next
    next.clear();
```

```
std::vector<int> frontier, next;
bool isFound = false
#pragma omp parallel
while (!isFound && frontier.size() > 0):
#pragma omp for nowait
    for offset in frontier:
        node = read_node(offset)
        if (node.key == key) isFound = true
        if (isFound || node's children count == 0)
        continue;
#pragma omp critical
    next.insert(node.children);
    frontier = next
    next.clear();
```


## Code

## Serial and Parallel DFS:

```
std::vector<int> frontier;
bool isFound = false;
frontier.push_back(0);
while (!frontier.empty()):
    node = read_node(frontier.back());
    if (node.key == key) isFound = true
    frontier.pop_back();
    frontier.insert(node.children);
return isFound;
```

```
std::vector<int> frontier;
bool isFound = false
frontier.push_back(0);
#pragma omp taskgroup
while (!frontier.empty()):
    node = read_node(frontier.back());
    if (node.key == key) isFound = true;
    frontier.pop_back();
    frontier.insert(node.children);
    while (frontier.size() > fSize):
        frontier, frontier_new = frontier.split();
#pragma omp task shared(isFound)
    if (run(frontier_new)){
        isFound = true;
#pragma omp cancel taskgroup
return isFound;
```

We also implement IDDFS and some hybrid approach, see
code if interested.

## Tree Structure

- Our testing was conducted on a randomized tree-like graph, with control over its branching factor, the branching factor of the individual nodes, and the number of values allowed per key
- This offers significant flexibility is studying configurations that are beneficial for concurrent search performance
- This also allows us to test worst and best-case scenarios for our search algorithms


## File Structure

- Memory-mapped structs allow efficient byte-level access of all the data within each node of tree
- Tree is serialized in BFS order, converting each node encountered into an S_Node, a node format that allows for efficient concurrent access of a node and it's children
- S_Node structs store the offsets of the children nodes in the file, making the file compatible with both DFS and BFS-based search schemes ${ }^{1}$
- Low-level syscalls allow us to avoid OS intervention and get more accurate results vs. Standard Library


## Serializer and Operators

- File is opened via the open(2) syscall (or CreateFileA on Windows) with the appropriate flags (see Direct I/O)
- The file descriptor is stored in the serializer object for reading from and writing to the file.
- Operators:
- S_Node* readNode(): reads node at current fd position
- S_Node* readNodefromOffset(size_t offset): reads node at position offset bytes forward relative to start of file
- void writeNodeWithOffset(size_t offset): writes node at position offset bytes forward relative to start of file
- void write_offset_metadata(): unused
- void read_offset_metadata(): unused
struct S_Node \{
int key;
int numChildren;
int payload[8];
int children[8];
\}
__attribute((aligned(512)));
struct Node \{
size_t numChildren;
size_t maxChildren; size_t numValues; size_t maxValues; int values[8]; Node* children[8];


## File I/O

No caching: Test concurrency of storage
Concurrent: I/O operations should not be serialized by OS somehow
No universal libs for now, have to use system calls.

- SSD device which supports concurrent I/O
- O_DIRECT, pread/pwrite (Linux)
- Windows: FILE_FLAG_NO_BUFFERING, FILE_FLAG_OVERLAPPED, GetOverlappedResult()
- Read/write with offsets


## Experiments - Direct I/O (Random Read)



Experimental Setting: 8-CPU
omp_set_num_threads

Summary The omp_set_num_threads routine affects the number of threads to be used for subsequent parallel regions that do not specify a num_threads clause, by setting the value of the first element of the nthreads-var ICV of the current task

## Format

C/C++
void omp_set_num_threads (int num_threads);
$\qquad$

## Experiments - Tuning fSize (Target Key Not Exist)



Experimental Setting:

- Num of Nodes: 20000
- Branch Size: 8
- Target Key Not Exist
- 8 CPU


## Experiments - Tuning fSize (Random Target Key)



Experimental Setting:

- Num of Nodes: 20000/50000/80000
- Branch Size: 2/4/8
- 20 Random Target Key +1 Not Exist
- 8 CPU \& 32 Threads


## Experiments - Algorithm Efficiency

## Experimental Setting:

- 20 Random Target Key +1 Not Exist
- 8 CPU \& 32 Threads



## Future Works

- Algorithm
- Better controls over scheduling and task stealing
- Hybrid search, adjust parameters automatically
- Extend algorithm to general graph structures
- Testing
- More different workload
- I/O Matrices like IOPS


## Some Lessons!

- Try and choose way of implementation wisely.
- Go: chan, operator(<-,->)
- C/C++: Std::thread, OpenMP
- Test in correct way.
- All tests in memory before midterm, no performance gain, puzzled.
- Leave time for debugging!
- especially for concurrent programming...


## Thanks!

## Reference

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