

Datastructures and Algorithms Greedy Algorithms, Huffman Coding (Trees), Parallel Programming

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Overview

Learning Objectives Huffman Coding Greedy Choice In-Class-Exercise (practical) Parallel Programming Old Exam Questions Hints for current tasks



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1. Learning Objectives

Objectives

- □ Be able to build a Huffman Coding Tree using the algorithm outlined in the session
- □ Be able to reason about simple multithreaded programs
- □ Understand the different aproaches to modelling performance of parallel programs (Amdahl, Gustafson)

2. Summary

Getting on the same page

■ What did you see in the lectures up to now?

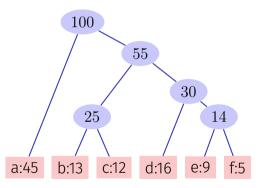
3. Huffman Coding

From the Lecture

Huffman's Idea

Tree construction bottom up

- Start with the set *C* of code words
- Replace iteratively the two nodes with _____ frequency by a _____. Replace iteratively the two nodes with smallest frequency by a new parent node.



Algorithm Huffman(C)

From the Lecture

// extract word with minimal frequency.

return ExtractMin(Q)

lnsert(Q, z)

4. Greedy Choice

Recap: Greedy Choice

Question:

What properties must an optimization problem with a recursive solution have in order to be solvable with a greedy algorithm? Also, give an example and a counterexample.

Recap: Greedy Choice

A problem with a recursive solution can be solved with a **greedy algorithm** if it has the following properties:

- The problem has optimal substructure: the solution of a problem can be constructed with a combination of solutions of sub-problems.
- The problem has the greedy choice property: The solution to a problem can be constructed, by using a local property that does not depend on the solution of the sub-problems.

Examples: Fractional knapsack problem, Huffman coding Counterexamples: Knapsack problem, optimal binary search tree.

5. In-Class-Exercise (practical)

Complement the DP implementation to compute an optimal search tree. \longrightarrow CodeExpert



6. Parallel Programming

Parallel Programming

Parallel Programming = perform multiple computations in parallel

Some terminology

Tasks

are computations that need to be done. Independent computations can be done in parallel.

Threads

are parallel executions, that execute tasks.

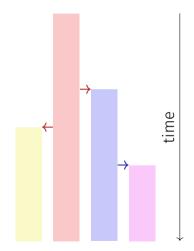
Shared resources

anything that is needed to perform tasks, but must be shared because there isn't a resource per task. (Not the focus for this week)

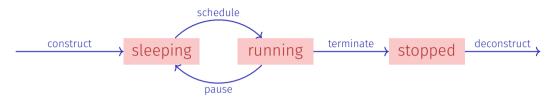
Forking Threads

Forking a thread means starting a new, concurrent computation

- Main thread forks a new thread
- Forking is done by creating a new thread object std::thread(func, args...)
- Main thread is the parent of its child thread
- Each thread can fork further threads



Thread Lifecycle (simplified)



The operating system's **scheduler** decides

- which thread can execute next (schedule)
- on which core to execute
- when to pause/sleep again

Switching threads on the processor, which puts the current thread to sleep and wakes up another one, is called **context switching**.

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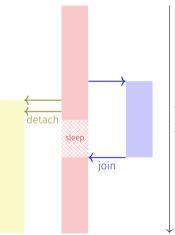
Join and Detach

other_thread.join() means waiting until
other_thread has finished execution.

 The joining thread will sleep until other_thread terminated (if it already did, no sleeping is necessary)

other_thread.detach() indicates that no thread
will wait for other_thread to finish execution

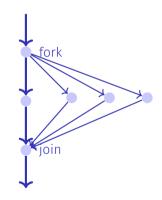
- Useful for nonterminating processes (e.g. servers), and reactive systems (e.g. GUIs)
- Terminates alongside the main thread at the latest (int main())



C++ Threads

```
void hello(unsigned id) {
  std::cout << "hello from " << id << "\n";
}</pre>
```

```
int main() {
  std::vector<std::thread> tv(3);
  unsigned id = 0;
  for (auto& t : tv)
    t = std::thread(hello, ++id);
  std::cout << "hello from main\n";
  for (auto& t : tv)
    t.join();
}</pre>
```



From the

Lecture

Nondeterministic Execution!

One execution:

hello from main hello from 1 hello from 2 hello from 3 Other execution: hello from 2 hello from main hello from 1 hello from 3 Other execution: hello from main hello from 1 hello from hello from 2 3

Technical Details I

Forking a function that takes a reference requires std::ref upon thread construction

```
void calc(std::vector<int>& very_long_vector) {
```

```
// doing funky stuff with very_long_vector
}
```

```
// main
std::vector<int> v(100000000);
```

```
std::thread t1(calc, std::ref(v)); // Compiler error w/o std::ref
std::thread t2([&v]{ calc(v)}; }); // Alternative
```

Technical Details II

Threads cannot be copied

```
// --- Error ---
std::thread t1(hello);
std::thread t2;
t2 = t1; // Compiler error
t1.join();
```

```
// --- OK ---
std::thread t1(hello);
std::thread t2;
t2 = std::move(t1); // OK
t2.join();
```

Also relevant if threads are to be stored in containers

Technical Details

Also see the corresponding "Exercise Class Example" on Code Expert with further technical details

Quiz

void print(char c); // Output character c

```
void A(char value) {
  if (value != 'D') {
   std::thread t(A, value + 1);
   print(value);
   t.join();
 }
}
int main() {
 std::thread t(A,'A');
 t.join();
}
```

possible output(s)?

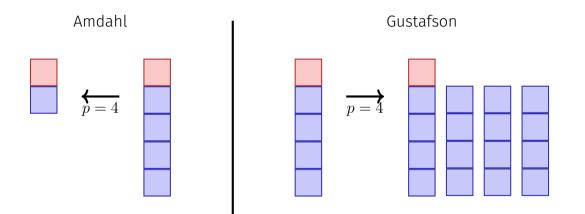
Parallel Performance

Given

- fixed amount of computing work W (number computing steps)
- Sequential execution time T_1
- **Parallel execution time on** p CPUs T_p

	runtime	speedup	efficiency
perfection (linear)	$T_p = T_1/p$	$S_p = p$	$E_p = 1$
loss (sublinear)	$T_p > T_1/p$	$S_p < p$	$E_p < 1$
sorcery (superlinear)	$T_p < T_1/p$	$S_p > p$	$E_p > 1$

Amdahl vs. Gustafson



Amdahl vs. Gustafson, or why do we care?

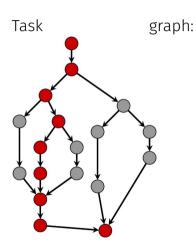
Amdahl | Gustafson pessimist | optimist strong scaling | weak scaling

 \Rightarrow need to develop methods with smallest sequential protion possible.

Performance Model

- *T*₁: **work**: time for executing total work on one processor
- **\blacksquare** T_p : Execution time on p processors
- T_∞: span: critical path, execution time on ∞ processors. Longest path from root to sink.
- T_1/T_∞ : **Parallelism:** wider is better
- Lower bounds:

$$T_p \geq T_1/p$$
 Work law $T_p \geq T_\infty$ Span law



From the

Lecture

Greedy scheduler: at each time it schedules as many available tasks as possible.

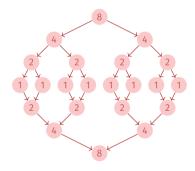
Theorem 1

On an ideal parallel computer with p processors, a greedy scheduler executes a multi-threaded computation with work T_1 and span T_∞ in time

 $T_p \le T_1/p + T_\infty$

Quiz: Scheduling

The following figure shows a task-graph of some algorithm. The number in each of the nodes denotes the execution time per task step.



$$T_{\infty} = ? \qquad T_1 = ? \qquad T_4 \le ? \qquad T_8 \le ?$$

7. Old Exam Questions

Old Exam Questions

Question

The analysis of a program has shown a speed-up of 2 when running on 9 processor cores. What is the serial fraction according to Gustafson's law?

Answer

Using Gustafson's law formula $S_p = p - \lambda \cdot (p - 1)$, we substitute the given values $S_p = 2$ and p = 9 to get $2 = 9 - \lambda \cdot 8$. Rearranging gives $7 = \lambda \cdot 8$. Solving for λ (the serial fraction), we find $\lambda = \frac{7}{8} = 0.875$.

Old Exam Questions

Question

You make a measurement of your program using a very large number of processor cores. The measurements suggest that the speed-up (using arbitrarily many processor cores) is bounded from above by $S_{\infty} = 2.5$. What is the best possible upper bound on the speed-up using 6 cores, assuming that Amdahl's law holds for your problem?

Answer

Using Amdahl's law formula $S_p \leq \frac{1}{\lambda + \frac{1-\lambda}{p}}$ and $S_{\infty} = \frac{1}{\lambda} = \frac{5}{2}$, we find $\lambda = \frac{2}{5}$. Substituting λ and p = 6 into Amdahl's law gives $S_6 \leq \frac{1}{\frac{0.4}{6} + \frac{0.6}{6}} = 2$.

8. Hints for current tasks

Huffman Coding

Huffman: Frequencies

```
Use std::unordered map (#include <unordered map>)
std::unordered_map<char, int> frequencies;
// ...
++frequencies['a'];
++frequencies['x'];
++frequencies['a'];
// A map is a container of key-value pairs (std::pair).
// Output all entries:
for (auto x:observations){
  std::cout << "observations of " << x.first << ":" << x.second << '\n';</pre>
}
```

Huffman: Min Heap

```
Use std::priority queue (#include <queue>)
struct MyClass {
  int x:
 MyClass(int X): x{X} {}
};
struct compare {
 bool operator() (const MyClass& a, const MyClass& b) const {
   return a.x < b.x:
 }
}:
std::priority queue<MyClass, std::vector<MyClass>, compare> q;
q.push(MyClass(10));
```

Huffman: Shared Pointers [optional]

Shared Pointers std::shared_ptr (#include <memory>)

```
struct SNode {
    int value;
    std::shared_ptr<SNode> left;
    std::shared_ptr<SNode> right;
    SNode(int v): value{v}, left{nullptr}, right{nullptr} {};
};
```

```
// A graph in which node 7 is shared: // 0
SNode* root = new SNode(0); // / \
root->left = new SNode(1); // 1 2
root->right = new SNode(2); // / \
root->right->left = new SNode(7); // \
root->right->right = root->right->left; // 7
```

root->left = nullptr; // Node 1 can and should be deallocated (deleted) now root->right->left = nullptr; // Node 7 must not yet be deallocated root->right->right = nullptr; // Node 7 can and should be deallocated now

Automated memory management, see Code Expert example

Huffman: Tree Nodes

```
using SharedNode = std::shared_ptr<Node>;
struct Node {
 char value;
 int frequency;
 SharedNode left:
 SharedNode right;
 // constructor for leafs
 Node(char v, int f):
   value{v}, frequency{f}, left{nullptr}, right{nullptr}
 {}
 // constructor for inner nodes
 Node(SharedNode 1, SharedNode r):
   value{0}, frequency{l->frequency + r->frequency}, left{l}, right{r}
 {}
}:
```

9. Outro

General Questions?

See you next time!

Have a nice week!