INF102 Algorithms and Data Structures

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INF102

- ► Lecturer: Marc Bezem, teaching assistants: NN
- ► Homepage: INF102 (hyperlinks in red)
- ► Textbook: Algorithms, 4th edition, R. Sedgewick and K. Wayne, Pearson, 2011
- ▶ Prerequisites: INF100 + 101 (\approx Ch. 1.1 + 1.2)
- Syllabus (pensum): Ch. 1.3–1.5, Ch. 2, Ch. 3, Ch. 4
- Exam: two or three compulsory exercises and a written exam
- ▶ Old exams: 2004–2013, 2014
- Contents of these slides here

Didactical stuff

- ► Good textbook from USA: many pages, exercises etc.
- Average speed must be ca 50 pages p/w
- Lectures focus on the essentials
- Prepare yourself by reading in advance
- Workshops about selected exercises
- ► Test yourself by trying some exercises in advance
- ▶ If you can do the exercises (incl. compulsory), you are fine

Generic Bags, Queues and Stacks

- ► Generic programming in Java, example: PolyPair.java
- ▶ Bag, Queue and Stack are generic, iterable collections
- Queue and Stack: Ch. 9 in textbook INF100/1
- ► APIs include: boolean isEmpty() and int size()
- All three support adding an element
- Queue and Stack support removing an element (if any)
- FIFO Queue, LIFO Stack
- Dijkstra's Two-Stack Expression Evaluation Movie

Implementations

- ResizingArray_Stack.java
- Resizing takes time and space proportional to size
- LinkedList_Stack.java
- Pointers take space and dereferencing takes time
- ▶ Programming with pointers: make a picture
- LinkedList_Queue.java

Computation time and memory space

- Two central questions:
 - ► How long will my program take?
 - Will there be enough memory?
- Example: TheeSum.java
- Inner loop is important

Methods of Analysis

Empirical:

- ▶ Run program with randomized inputs, measuring time & space
- Run program repeatedly, doubling the input size
- Measuring time: StopWatch
- Plot, or log-log plot and linear regression

Theoretical:

- Define a cost model by abstraction (e.g., array accesses, comparisons, operations)
- Try to count/estimate/average this cost as function of the input (size)
- ▶ Use O(f(n)) and $f(n) \sim g(n)$

ThreeSum, empirically

- ▶ Input sizes 1K, 2K, 4K, 8K take time 0.1, 0.8, 6.4 ,51.1 sec
- ► The log's are 3, 3.3, 3.6, 3.9 and -1, -0.1, 0.8, 1.71
- ▶ Linear regression gives $y \approx 3x 10$
- ▶ $\lg(f(n)) = 3\lg(n) 10$ iff

$$f(n) = 10^{\lg(f(n))} = 10^{3\lg(n)-10} = n^3 * 10^{-10}$$

- ▶ Conclusion: cubic in the input size, with constant $\approx 10^{-10}$
- Strong dependence on input can be a problem
- ightharpoonup Constant 10^{-10} depends on computer, exponent 3 does not

ThreeSum, theoretically

- ▶ Number of different picks of triples: g(n) = n(n-1)(n-2)/6
- ▶ Inner loop executed g(n) times
- $g(n) = n^3/6 n^2/2 + n/3$
- ▶ Cubic term $n^3/6$ wins for large n
- ▶ Computational model # array accesses: $n^3/2$
- ► Cost array access t sec: time $t * n^3/2$ sec
- Cost models are abstractions! (NB cache)

Big Oh, and \sim

- Q: 'wins for large n' uhh???
- lacktriangle A: Big Oh, and \sim will clear this up
- ▶ Costs are positive quantities, so $f, g, ... : \mathbb{N} \to \mathbb{R}^+$
- ▶ MNF130: f(n) is O(g(n)) if there exist c, N such that $f(n) \le cg(n)$ for all $n \ge N$
- ► Example: n^2 and even $99n^3$ are $O(n^3)$, but n^3 is not $O(n^{2.9})$
- ▶ INF102: $f(n) \sim g(n)$ if $1 = \lim_{n \to \infty} f(n)/g(n)$
- ▶ If $f(n) \sim g(n)$, then f(n) is O(g(n)) and g(n) is O(f(n))
- ▶ Big Oh and ~ aim to capture 'order of growth'
- ightharpoonup Big Oh abstracts from constant factors, \sim does not
- Large constant factors are important!

Important orders of growth

- ▶ constant: c(f(n) = c for all n)
- ▶ linear: n (compare all for n = 20 sec)
- ▶ linearithmetic: n lg n
- ▶ quadratic: n²
- ightharpoonup cubic: n^3
- exponential: 2ⁿ
- general form: $an^b(\lg n)^c$

Examples

- Worst case: guaranteed, independent of input
- ► Average case: not guaranteed, dependent of input distribution
- Linked list implementations of Stack, Queue and Bag: all operations take constant time in the worst case
- Resizing array implementations of Stack, Queue and Bag: adding and deleting take linear time in the worst case (easy)
- Resizing array implementations of Stack, Queue and Bag: adding and deleting take on average constant time in the worst case (difficult)
- Special case of resizing array that is only growing: $1(2)2(4)34(8)5678(16)9 \dots 16(32) \dots$, with (n) the new size. Risizing to (n) costs 2n array accesses, so in total $(1+4)+(1+8)+(2+16)+(4+32)+(8+64) \dots$, so 9 per push.

Staying Connected

- ▶ MNF130: relation $R \subseteq V \times V$ is an *equivalence* if
 - ▶ R is reflexive: $\forall x \in V$. R(x,x)
 - ▶ *R* is *symmetic*: $\forall x, y \in V$. $R(x, y) \rightarrow R(y, x)$
 - ▶ R is transitive: $\forall x, y, z \in V$. $R(x, y) \land R(y, z) \rightarrow R(x, z)$
- We assume connectedness to be an equivalence
- Dynamic connectivity means that R can grow and shrink
- ► Example: if the 'Bergensbanen' is broken, Oslo and Bergen are no longer connected by rail
- We want efficient algorithms and datastructures for testing whether two objects are connected
- Clear relationship with paths in graphs, more in Ch. 4
- ▶ Here we take $V = \{0, ..., N 1\}$.

Staying Connected

- ▶ MNF130: relation $E \subseteq V \times V$ is an *equivalence* if
 - ▶ *E* is reflexive: $\forall x \in V$. E(x,x)
 - ▶ *E* is *symmetic*: $\forall x, y \in V$. $E(x, y) \rightarrow E(y, x)$
 - ▶ *E* is transitive: $\forall x, y, z \in V$. $E(x, y) \land E(y, z) \rightarrow E(x, z)$
- We assume connectedness to be an equivalence
- Dynamic connectivity means that R can grow and shrink
- ► Example: if the 'Bergensbanen' is broken, Oslo and Bergen are no longer connected by rail
- We want efficient algorithms and datastructures for testing whether two objects are connected
- Clear relationship with paths in graphs, (connected) components (MNF130)
- We take $V = \{0, ..., N-1\}$.

Union Find

- ▶ UF, idea: every component has an identifier ('hub'), which has edges ('spokes') to the elements of its component
- ► API: UF
- ▶ Implementations with int[] id containing the identifiers
 - ► SlowUF.java
 - ► FastUF.java
 - WeightedUF.java
- WeightedUF: log depth of tree (Proposition X)

Sorting

- Sorting: putting objects in a certain order
- ▶ MNF130: relation $R \subseteq V \times V$ is a total order(ing) if
 - 1. R is reflexive: $\forall x \in V$. R(x,x)
 - 2. R is transitive: $\forall x, y, z \in V$. $R(x, y) \land R(y, z) \rightarrow R(x, z)$
 - 3. R is antisymmetric: $\forall x, y \in V$. $R(x, y) \land R(y, x) \rightarrow x = y$
 - 4. R is total: $\forall x, y \in V$. $R(x, y) \vee R(y, x)$
- Natural orderings:
 - ▶ Numbers of any type: ordinary \leq and \geq
 - ► Strings: lexicographic
 - ► Objects of a Comparable type: v.compareTo(w) < 0

Sorting (ctnd)

- Bubblesort: ExampleSort.java
- Certification: assert isSorted(a) in main()
- No guarantee against modifying the array (but exch() is safe)
- Costmodel 1: number of exch()'s and less()'s
- Costmodel 2: number of array accesses
- ► Pitfalls: cache misses, expensive v.compareTo(w) < 0
- Why studying sorting? (java.util.Arrays.sort())
- Comparing sorting algorithms: CompareSort.java

ToC and topics of general interest

- ► Table of Contents on next slide (all items clickable)
- Practical stuff: slide 2

Introduction

Ch.1.3 Bags, Queues and Stacks

Ch.1.4 Analysis of Algorithms

Ch.1.5 Case Study: Union-Find

Ch.2.1 Elementary Sorts

Ch.2.2 Mergesort

Ch.2.3 Quicksort

Ch.2.4 Priority Queues

Ch.3.1 Symbol Tables

Ch.3.2 Binary Search Trees

Ch.3.3 BalancesSearch Trees

Ch.3.4 Hash Tables

Ch.3.5 Applicatios

Ch.4.1 Undirected Graphs

Ch.4.2 Directed Graphs

Ch.4.3 Minimum Spanning

Trees

Ch.4.4 Shortest Paths

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