

Verified Algorithm Analysis: Correctness and Complexity

A Biased Survey

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Focus on algorithm analyses in ITPs

Unless otherwise noted: in Isabelle/HOL

Please let me know of missed references

Out of scope: related work on completely automatic running time analyses by Martin Hofmann, Jan Hoffmann, Madhavan & Kuncak, ...

- ① Mathematical Foundations
- ② Programming and Verification Frameworks
- ③ Algorithms

① Mathematical Foundations

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Slides and results by Manuel Eberl

Classic concepts and results

- Landau symbols
- Generating functions
- Linear recurrences (theory and solver)
- Asymptotics of $n!$, Γ , H_n , C_n , \dots

Akra–Bazzi theorem

Generalisation of the
Master Theorem for divide-and-conquer recurrences

Input (simple case):

$$T(x) = g(x) + \sum_{i=1}^k a_i T(\lfloor b_i x \rfloor) \quad \text{for } g \in \Theta(x^q \ln^r x)$$

Result:

$$\begin{array}{ll} T \in \Theta(x^p) & T \in \Theta(x^p \ln \ln x) \\ T \in \Theta(x^q) & T \in \Theta(x^p \ln^{q+1} x) \end{array}$$

where p is the unique solution to $\sum a_i b_i^p = 1$

Examples for Akra–Bazzi

| Algorithm | Recurrence | Solution |
|-----------------|--|-------------------|
| Binary search | $T(\lceil n/2 \rceil) + O(1)$ | $O(\log n)$ |
| Merge sort | $T(\lfloor n/2 \rfloor) + T(\lceil n/2 \rceil) + O(n)$ | $O(n \log n)$ |
| Karatsuba | $2T(\lceil n/2 \rceil) + T(\lfloor n/2 \rfloor) + O(n)$ | $O(n^{\log_2 3})$ |
| Median-of-med's | $T(\lceil 0.2n \rceil) + T(\lceil 0.7n \rceil + 6) + O(n)$ | $O(n)$ |

All of this is (almost) automatic.

Automated asymptotics

Isabelle can automatically prove

- $f(x) \xrightarrow{x \rightarrow L} L'$
- $f \in O(g)$, $f \in o(g)$, $f \in \Theta(g)$, $f(x) \sim g(x)$
- $f(x) \leq g(x)$ for x sufficiently close to L

for a wide class of \mathbb{R} -valued functions/sequences.

How? Multiseries expansions

Similar to algorithms used in Mathematica/Maple

Automated asymptotics

Example from Akra–Bazzi proof:

$$\lim_{x \rightarrow \infty} \left(1 - \frac{1}{b \log^{1+\varepsilon} x} \right)^p \left(1 + \frac{1}{\log^{\varepsilon/2} \left(bx + \frac{x}{\log^{1+\varepsilon} x} \right)} \right) - \left(1 + \frac{1}{\log^{\varepsilon/2} x} \right) = 0^+$$

Can be proved automatically in 0.3 s.

① Mathematical Foundations

② Programming and Verification Frameworks

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For programming, refinement and verification
of algorithms
in Isabelle/HOL

Functional vs Imperative

Functional algorithms are expressed as HOL functions

Imperative algorithms are expressed in

Imperative HOL

a monadic framework with arrays and references by
Bulwahn & Co [TPHOLs 08]

Can generate code in SML, OCaml, Haskell and Scala
[Haftmann, N. FLOPS 10]

A problem:

Head-on verification of efficient algorithms
is painful or impossible

The cure:

Start from an abstract functional version
and refine it to an efficient (imperative) algorithm

A second problem:

Not every algorithm is deterministic:
for every neighbour do ...

Isabelle refinement framework

Lammich [ITP 12, ITP 13, ITP 15, CPP 16]

Provides abstract programming language with

- nondeterminism
- loops (incl. `foreach`)
- general recursion
- specification statement

Isabelle refinement framework

Lammich [ITP 12, ITP 13, ITP 15, CPP 16]

Stepwise program refinement by:

- algorithm refinement
- semi-automatic data refinement
using verified collections library
- semi-automatic refinement to Imperative HOL

Almost all referenced Isabelle proofs can be found in the
[Archive of Formal Proofs \(AFP\)](#)

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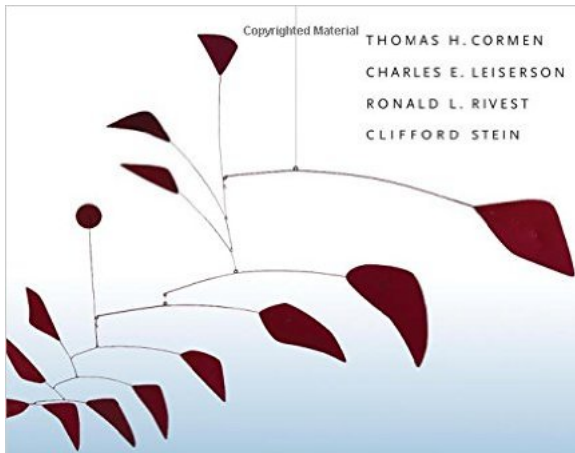
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INTRODUCTION TO

ALGORITHMS

THIRD EDITION

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③ Algorithms

Sorting & Order statistics

Search trees

Advanced Design and Analysis Techniques

Dynamic Programming

Advanced Data Structures

Graph Algorithms

Randomized Algorithms

Sorting

TIMsort: `java.util.Arrays.sort`

- A complex combination of mergesort and insertion sort on arrays
- De Gouw & Co [[CAV 15](#)] discover bug and suggest fixes
- De Gouw & Co [[JAR 17](#)] verify termination and exception freedom using the KeY system.
Meanwhile: verification of functional correctness

k-th smallest element via median of medians

```
select k xs =  
  let x = select ... (map median5 (chop 5 xs));  
      (ls, es, gs) = partition3 x xs  
  in if ... then select k ls  
     else ... select ... gs
```

- Functional version by Eberl [[AFP 17](#)]
- Imperative refinement (incl linear time proof)
by [Zhan](#) & [Haslbeck](#) [[IJCAR 18](#)] using Akra-Bazzi

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Popular case study for ITPs because nicely functional.

AVL and Red-Black trees:

- Filliâtre & Letouzey [ESOP 04] (in Coq)
- N. & Pusch [AFP 04]
- Krauss & Reiter [08]
- Charguéraud [10] (in Coq)
- Appel [11] (in Coq)
- Dross & Moy [14] (in SPARK)
- ...

Functional correctness

- Functional correctness obvious to humans
but until recently more or less verbose in ITPs
- Most verifications based on some variant of $bst\langle l, a, r \rangle \leftrightarrow bst\ l \wedge bst\ r \wedge (\forall x \in l. x < a) \wedge (\forall x \in r. a < x)$
- Correctness proofs can be automated if $bst(t)$ is replaced by N. $sorted(inorder\ t)$ [N. ITP 16]
- Works for AVL, RBT, 2-3, 2-3-4, AA, splay and other search trees covered in this talk
- Not automated: balance invariants

Some more search trees
Not in CLRS

Weight-Balanced Trees

Nievergelt & Reingold [[72,73](#)]

- Parameter: balance factor $0 < \alpha \leq 0.5$
- Every subtree must be balanced:

$$\alpha \leq \frac{\text{size of smaller child}}{\text{size of whole subtree}}$$

- Insertion and deletion: single and double rotations depending on subtle numeric conditions
- Nievergelt and Reingold deletion incorrect
- Mistake discovered and corrected by Blum & Mehlhorn [[80](#)] and Hirai & Yamamoto [[JFP 11](#)] (in Coq)

Scapegoat trees

Anderson [89,99], Igal & Rivest [93]

Central idea:

Don't rebalance every time,
Rebuild when the tree gets “too unbalanced”

- Tricky: amortized logarithmic complexity analysis
- Recently verified [N. APLAS 17]

Functional finger tree

Hinze & Paterson [06]

Tree representation of sequences with

- access time to both ends in amortized $O(1)$
- concatenation and splitting in $O(\log n)$

General purpose data structure for implementing *sequences, priority queues, search trees, ...*

Verifications:

- Functional correctness:
 - Sozeau [ICFP 07] (in Coq)
 - Nordhoff, Körner, Lammich [AFP 10]
- Amortized complexity:
 - Danielsson [POPL 08] (in Agda)

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Huffman's algorithm

Huffman [52]

- Purpose: lossless text compression, eg Unix zip command
- Input: frequency table for all characters
- Output:
variable length *binary code* for each character
that minimizes the length of the encoded text
⇒ short codes for frequent characters
- Functional correctness proof: Blanchette [JAR 09]

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The functional approach

Wimmer & Co [ITP 18]

Write recursive program

```
fib(n) = fib(n-1) + fib(n-2)
```

Crank the handle and obtain monadic memoized version

```
fib' n := do { a ← fib'(n-1);  
              b ← fib'(n-2);  
              return (a+b) }
```

with correctness theorem

```
snd (runstate (fib' n) empty) = fib n
```

where `f x := rhs` abbreviates

```
f x = do a ← lookup x;  
      case a of  
        Some r ⇒ return r |  
        None ⇒ do r ← rhs;  
              update x r;  
              return r
```

Automation

- Automatic definition of monadic memoized function
- Automatic correctness proof via parametricity reasoning

How is the state (= memory) realized?

Two state monads

- Purely functional state monad based on some search tree
- State monad of Imperative HOL using arrays
Same $O(\cdot)$ running time as standard imperative programs

Applications

- Bellman-Ford (SSSP)
- CYK (Context-free parsing)
- Minimum Edit Distance
- Optimal Binary Search Tree
- ...

Including correctness proofs

But without complexity analysis (yet)

Optimal Binary Search Tree

Input:

- set of keys k_1, \dots, k_n
- access frequencies b_1, \dots, b_n (hits):
 $b_i =$ number of searches for k_i
- and a_0, \dots, a_n (misses):
 $a_i =$ number of searches in (k_i, k_{i+1})

Algorithms for building optimal search tree:

- Straightforward recursive cubic algorithm
- Knuth [71]: a quadratic optimization
- Yao [80]: simpler proof
- N. & Somogyi [AFP 18]

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B-trees

Functional verification:

- Malecha & Co [POPL 10] (in Coq + Ynot)
- Ernst & Co [SSM 15] (in KIV)

Priority queues

Verification of functional implementations:

- Leftist heap
- Braun tree [N. [AFP 14](#)]
- Amortized analysis of
Skew heap, Splay heap, Pairing heap
N. [[ITP 16](#)], N. & Brinkop [[JAR 18](#)]
- Binomial heap and **Skew binomial heap**
Meis, Nielsen, Lammich [[AFP 10](#)]

None of the above provide decrease-key ...

Challenge!

Union-Find

Charguéraud, Pottier, Guéneau [[ITP 15](#), [JAR 17](#), [ESOP 18](#)]

Framework (“Characteristic Formula”):

- Translates OCaml program into a logical formula that captures the program behaviour, including effects and running time.
- Import into Coq as axiom
- Verify program in Coq

Verified amortized complexity $O(\alpha(n))$ of each call
(Following Alstrup & Co [JA 14])

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Strongly connected components

- Tarjan [72],
verified by Schimpf & Smaus [ICLA 2015]
- Gabow [IPL 00],
verified by Lammich [ITP 14]

Used in verified model checker CAVA

Dijkstra (SSSP)

Dijkstra [59]

Functional correctness verified:

- Nordhoff & Lammich [AFP 12]:
purely functionally with finger trees
- Lammich [CPP 16]:
imperative with arrays

Floyd-Warshall (APSP)

Functional correctness verified by
Wimmer & Lammich [AFP 17]:

- Functional implementation
- Refined to imperative algorithm on an array
- Main complication: destructive update
- All related verifications make simplifying assumptions — also in CLRS

Maximum network flow

- Edmonds-Karp:
Lammich & Sefidgar [ITP 16]
Imperative, running time $O(|V||E|^2)$
- Push-Relabel (2 variants):
Lammich & Sefidgar [JAR 17]
Imperative, running time $O(|V|^2|E|)$

Competitive with a Java implementation

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Randomized algorithms formalized

Purely functionally via the Giriy monad

Example:

```
do { a ← some distribution;  
    b ← some other distribution (a);  
    return (a+b) }
```

Quicksort

- van der Weegen & McKinna [ITP 08] (in Coq)
Proved expected running time of randomized and deterministic quicksort $\leq 2n \lceil \log_2 n \rceil$
- Eberl & Co [ITP 18]
Proved closed form $2(n+1)H_n - 4n$
and asymptotics $\sim 2n \ln n$
- Tassarotti & Harper [ITP 18] (in Coq)
Formalized and extended cookbook method for tail bounds [Karp JACM 94]
Applied it to quicksort:
$$\Pr[T(n) > (c+1)n \log_{4/3} n + 1] \leq \frac{1}{n^{c-1}}$$

Analysis of random BSTs

Eberl & Co [ITP 18]

“Random BST” means

BST generated from a random permutation of keys

Thm Expected height of random BST

$$\leq \dots \sim 3 \log_2 n$$

Thm Distribution of internal path lengths

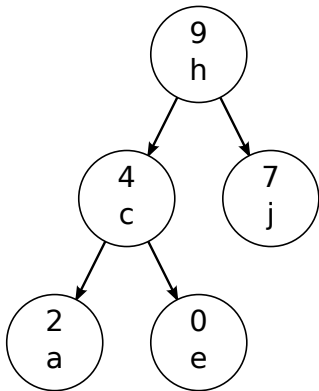
= distribution of running times of quicksort

Treaps

Aragon & Seidel [89, 96]

Random BSTs are pretty good,
but keys are typically not random

Treaps: combine each key with a random *priority*



treap = tree + heap

Treaps verified

Eberl & Co [ITP 18]

- Functional correctness straightforward
- Treaps need a *continuous* distribution of priorities to avoid duplicates (with probability 1)
- Reasoning about continuous distributions is hard because of measurability proofs
- **Thm** Distribution of treaps
= distribution of random BSTs (modulo priorities)

Conclusion: Comparison with CLRS

The first 750 pages (parts I–VI, the “core”)

- Much of the basic material has been verified
- Major omissions (afaik):
 - Hashing incl. probabilities
 - Fibonacci heaps
 - van Emde Boas trees