# Equational Reasoning 

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Exponential time because k requires $\log (\mathrm{k})$ bits to represent
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## Complexity Analysis

- With SAT, no one has come up with a polynomial time algorithm
- What about sum? Can we do better?
(definec fsum (k :nat) :nat What is the (/ (* k (+ k 1)) 2)) time complexity?

| Input (k) | Aritmetic Ops | Complexity (k) | Input Size (n) | Complexity (n) |
| :---: | :---: | :---: | :---: | :---: |
| 2 | 3 | $\mathrm{O}(1)$ | 1 | $\mathrm{O}(1)$ |
| 32 | 3 |  | 5 |  |
| 1024 | 3 | 10 |  |  |
| 32768 | 3 | 15 |  |  |
| 1048576 | 3 | 20 |  |  |

Constant time, so exponentially better than sum!
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## Reasoning About Arithmetic

(definec sum (k :nat) :nat (definec fsum (k :nat) :nat (if (= k 0) (/ (* k (+ k 1)) 2))
0
$(+k(\operatorname{sum}(-k 1)))))$

- We want to prove that a more clever version is equivalent (implies (natp k) (equal (sum k) (fsum k)) )
- How? By "mathematical induction" (think about 1800)

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Induction Proof

$$
\begin{aligned}
& \text { (definec sum (k :nat) :nat (definec fsum (k :nat) :nat } \\
& \text { (if (=k 0) } \\
& \text { (/ (* k (+ k 1)) 2)) } \\
& (+k(\operatorname{sum}(-k 1))))
\end{aligned}
$$

Conjecture: (natp $k) \Rightarrow($ sum $k)=($ fsum $k)$

- Base case:
$($ natp $k) \wedge k=0 \Rightarrow($ sum $k)=(1(* k k+1) 2)$
Induction step:

$$
\begin{aligned}
& (\text { natp } k) \wedge k \neq 0 \wedge \\
& {[(\text { natp } k-1) \Rightarrow(\text { sum } k-1)=(/(* k-1 k) 2)] } \\
\Rightarrow & (\text { sum } k)=(/(* k k+1) 2)
\end{aligned}
$$

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## ACL2s Demo

- Show that sum takes exponential time
- The importance of tail recursion
- fsum to the rescue

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## Lessons Learned

- Algorithmic complexity is vitally important: consider big-data, Web
- Take algorithms as soon as possible
- As a computer scientist, always think about complexity
- But, correctness is most important: fast, but wrong is not good
- Planes, trains and automobiles (not the movie) crash
- Wrong simulation results for weather, nuclear testing, experiments...
- Correctness is mostly what we care about in this class
- Powerful idea: define correctness using simplest definitions (the spec)
- Then define efficient implementation and prove equivalence
- Allows one to reason using the spec, but execute using efficient code

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## Comparison with C \& Java

- Suppose that we write this code in an imperative language like C or Java
- Let's see a DEMO
- What happened?

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## Limited Precision!

- C, Java, etc. do not have arbitrary precision arithmetic
- So sum, fsum are not equivalent!
- We get a negative number because most languages use fixed-bit arithmetic

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## Finding Bugs

- You could have tested your program 1K times and not found errors
- We knew what we were looking for and so we found an error
- Is this a problem in practice? Yes. See http:// googleresearch.blogspot.no/2006/06/extra-extra-read-all-about-itnearly.htm

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## Reasoning About C/Java

- Can we reason about C/Java code?
- We don't have a theorem prover for these languages
- But, we can reason about them!
- Use ACL2s to model arithmetic in C/Java
- Let's say that the spec is that fsum should be equal to sum
- We can use property-based testing
- DEMO

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