Automatización de Análisis Asintótico en Isabelle/HOL

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Agenda

- 1. A Smörgåsbord of Asymptotic Analysis in Isabelle/HOL
- 2. Formal Aspects of Asymptotics in Isabelle
- 3. Automating Asympotics

Disclaimer

I did not invent any of these things. I just figured out how to do them in Isabelle/HOL

A Smörgåsbord of Asymptotic Analysis in Isabelle/HOL

Relates the value of a sum to the corresponding integral

$$\sum_{i=a+1}^{b} f(i) - \int_{a}^{b} f(x) dx = \sum_{k=1}^{N} \frac{B_{k}}{k!} \left(f^{(k-1)}(b) - f^{(k-1)}(a) \right) + R$$

where $R = \frac{(-1)^{N}}{N!} \int_{a}^{b} \bar{B}_{N}(x) f(x) dx$

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Useful e.g. for n!, H_n , Γ , ψ , ...

Theory and solver for linear recurrences with constant coefficients, i. e.

$$a_0f(n) + \ldots + a_kf(n+k) = b_n$$
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- Average-case analysis of algorithms (cf. Flajolet)
- Analysis of probabilistic programs: Random Walk

The nuclear option for analysing asymptotics of Divide-and-Conquer recurrences

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Applications:

• Merge Sort: $f(\lfloor n/2 \rfloor) + f(\lceil n/2 \rceil) + n \implies \Theta(n \log n)$

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- Merge Sort: $f(\lfloor n/2 \rfloor) + f(\lceil n/2 \rceil) + n \Longrightarrow \Theta(n \log n)$
- QuickSelect: $f(\lfloor \frac{1}{5}n \rfloor) + f(\lfloor \frac{7}{10}n \rfloor + 6) + \frac{12}{5}n \Longrightarrow \Theta(n)$

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• Via Gamma function:
$$C_n \sim \frac{4^n}{\sqrt{\pi}n^{1.5}}$$

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$$B_n \sim 2(-1)^{n+1} \frac{(2n)!}{(2\pi)^{2n}}$$

Can also be used to prove $\zeta(2n) = \frac{1}{2}(-1)^{n+1}B_{2n}\frac{(2\pi)^{2n}}{(2n)!}$.

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- Other: General comparison sorting, QuickSort, BSTs, Treaps, Skip Lists

Formal Aspects of Asymptotics in Isabelle

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Filters also make it possible to say P(x) holds for all x sufficiently big / close to x_0 .

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 $qs_cost \in O[length going_to at_top](\lambda xs. (length xs)^2)$

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Things like $(\lambda x. 2 * x + x * \ln \ln x) \in O(\lambda x. x * \ln x)$ get proven automatically.

Automating Asymptotics

Example: Lemma required for Akra-Bazzi

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

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Original author: 'Trivial, just Taylor-expand it!' In Isabelle: 700 lines of messy proofs

lemma akra_bazzi_aux: filterlim $(\lambda x. (1 - 1/(b * \ln x^{(1 + \varepsilon)})^p) * (1 + \ln (b * x + x/\ln x^{(1 + \varepsilon)})^{(-\varepsilon/2)}) - (1 + \ln x^{(-\varepsilon/2)}))$ $(at_right 0) at_top$

$$\begin{array}{ll} \textbf{lemma akra_bazzi_aux:} \\ \textbf{filterlim} \\ & (\lambda x. \ (1 - 1/(b*\ln x \ (1 + \varepsilon)) \ p) * \\ & (1 + \ln \ (b*x + x/\ln x \ (1 + \varepsilon)) \ (-\varepsilon/2)) - \\ & (1 + \ln \ x \ (-\varepsilon/2))) \\ & (\textbf{at_right 0}) \ \textbf{at_top} \\ \textbf{by magic} \end{array}$$

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This is what we would like to have.

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This is what we would like to have.

Computer Algebra Systems can do this (sort of) So why can't we?

Related Work

- Asymptotic Expansions of exp-log Functions by Richardson, Salvy, Shackell, van der Hoeven
- On Computing Limits in a Symbolic Manipulation System by Gruntz

For $x \to \infty$, we have:

$$e^{1/x} \sim 1 + x^{-1} + \frac{1}{2}x^{-2} + \frac{1}{6}x^{-3} + \dots$$

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This means: Cutting off $f(x) \sim a_0(x) + a_1(x) + \ldots$ at term a_n yields error $O(a_{n+1}(x))$.

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Expansions contain the *full* asymptotic information. They can be added/subtracted/multiplied/divided. Limits can simply be 'read off'



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For now, we only consider expansions of the form

$$f(x) \sim c_0 x^{e_0} + c_1 x^{e_1} + \dots$$

for $x \to \infty$ where $e_0 > e_1 > \ldots$

How can one do concrete operations on these expansions? type Exp = $(\mathbb{R} \times \mathbb{R})$ llist

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 llist

negate : Exp
$$\rightarrow$$
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negate :
$$Exp \rightarrow Exp$$

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(+) : $Exp \rightarrow Exp \rightarrow Exp$
[] + $ys = ys$
 $xs + [] = xs$

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$$\begin{array}{l} \text{negate} : \, \text{Exp} \to \text{Exp} \\ \text{negate} \, xs \, = \, \left[(-c, e) \mid (c, e) \leftarrow xs \right] \\ (+) : \, \text{Exp} \to \text{Exp} \to \text{Exp} \\ \left[\right] \, + \, ys \, = \, ys \\ xs \, + \, \left[\right] \, = \, xs \\ ((c_1, e_1) :: \, xs) \, + \, ((c_2, e_2) :: \, ys) \\ \left| \, e_1 = = \, e_2 \, = \, (c_1 + c_2, e_1) \, :: \, xs + ys \\ \left| \, e_1 < e_2 \, = \, (c_1, e_1) \, :: \, xs + ((c_2, e_2) :: \, ys) \\ \left| \, e_1 > e_2 \, = \, (c_2, e_2) \, :: \, ((c_1, e_1) :: \, xs) + ys \end{array} \right. \end{array}$$
Asymptotic Expansions – Multiplication

Multiplication with 'atomic' factor $c'x^{e'}$:

$$\begin{array}{l} \mathsf{scale} \,:\, \mathbb{R} \to \mathbb{R} \to \mathsf{Exp} \to \mathsf{Exp} \\ \mathsf{scale} \,\, c' \,\, e' \,\, \mathsf{xs} \,=\, [(c \ast c', e + e') \mid (c, e) \leftarrow \mathsf{xs}] \end{array}$$

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Example: $(e^x, x, \ln x)$ is an asymptotic basis and generates monomials $e^{ax}x^b \ln^c x$

$$e^{4x} + 2x^3 \ln x \stackrel{\scriptscriptstyle \wedge}{=} [1 \cdot (4, 0, 0), 2 \cdot (0, 3, 1)]$$

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- Limit is 0 if $e_i < 0$
- Limit is $sgn(c) \cdot \infty$ if $e_i > 0$
- Limit is c if all $e_i = 0$

type $Exp = (\mathbb{R} \times \mathbb{R})$ llist negate : $Exp \rightarrow Exp$ negate $xs = [(-c, e) \mid (c, e) \leftarrow xs]$

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Now:

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type Basis = $(\mathbb{R} \to \mathbb{R})$ list **datatype** Exp : Basis \to Type where Const : $\mathbb{R} \to \text{Exp}$ [] Exp : $(\text{Exp } bs \times \mathbb{R})$ llist $\to \text{Exp } (b :: bs)$ negate : Exp $bs \to \text{Exp } bs$ negate (Const c) = -cnegate (Exp xs) = Exp [(negate c, e) | $(c, e) \leftarrow xs$]

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- Lots of opportunities for implementation bugs
- Luckily, the Isabelle kernel caught them, of course. :)

Skipping a lot of magic: We can automatically prove statements of the form

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sin, cos, tan at finite points also possible.

Example

lemma
$$(\lambda n. (1+1/n) \hat{n}) \longrightarrow \exp 1$$

by exp_log_asymptotics

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Example

lemma
$$((\lambda x. (1 + y/x)^x) \longrightarrow \exp y)$$
 at_top
proof (cases $y = 0$)
case False
thus ?thesis **by** exp_log_asymptotics
qed simp_all

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assumes c > 1 and k > 0shows $(\lambda n. n^k) \in o(\lambda n. c^n)$ using assms by exp_log_asymptotics

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Discussion

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- Case distinctions have to be done manually

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- Implementation was tricky to get right

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Nowadays, most of them work

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Maxima and Sage take very long for some of them and give wrong result for this:

$$\exp\left(\frac{\log\log\left(x+e^{\log x\log\log x}\right)}{\log\log\log\left(e^x+x+\ln x\right)}\right) \longrightarrow e$$

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But: you have to trust the implementations.

Isabelle still isn't a CAS – but we're getting there.

Future Work

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- Cannot handle oscillating functions
- User interaction for zeroness tests could be improved

Questions? Demo?