

# High-precision methods for zeta functions

## Part 2: derivatives

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# Computing derivatives

$$f(a+x) = \sum_{n=0}^{\infty} \frac{f^{(n)}(a)}{n!} x^n$$

Reasons we might want to compute derivatives

- ▶ We are interested in the derivatives / Taylor coefficients themselves
  - ▶ Removing singularities (computing limits)
  - ▶ Analytic operations (integrals, finding roots or extreme points).
- Newton's method:

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$$

# Numerical differentiation

Finite difference:

$$f^{(n)}(a) \approx \frac{1}{h^n} \sum_{k=0}^n (-1)^{k+n} \binom{n}{k} f(a + kh), \quad h \sim 2^{-p}$$

Cauchy integral formula:

$$f^{(n)}(a) = \frac{n!}{2\pi i} \oint_{\gamma} \frac{f(z)}{(z - a)^{n+1}} dz$$

## Symbolic differentiation

```
sage: diff(1/cos(x), x, 10)
```

```
50521/cos(x) + 1326122*sin(x)^2/cos(x)^3 +  
6749040*sin(x)^4/cos(x)^5 + 13335840*sin(x)^6/cos(x)^7 +  
11491200*sin(x)^8/cos(x)^9 + 3628800*sin(x)^10/cos(x)^11
```

$$\zeta^{(n)}(s) = ???$$

## Power series arithmetic (automatic differentiation)

We work with objects

$$f = f_0 + f_1x + \dots + f_{n-1}x^{n-1} \in \mathbb{C}[[x]]/\langle x^n \rangle$$

With  $n = 2$  (first derivatives):

$$(f_0 + f_1x) \times (g_0 + g_1x) = f_0g_0 + (f_0g_1 + f_1g_0)x$$

$$\frac{1}{f_0 + f_1x} = \frac{1}{f_0} - \frac{f_1}{f_0^2}x$$

$$\sin(f_0 + f_1x) = \sin(f_0) + \cos(f_1)x$$

## Formal and non-formal operations

$$\frac{1}{1-f} = 1 + f + f^2 + f^3 + \dots$$

With  $f = x + x^2$ :

$$1 + (x + x^2) + (x^2 + 2x^3 + x^4) + (x^3 + 3x^4 + 3x^5 + x^6) + \dots$$

With  $f = 0.5 + x$ :

$$1 + (0.5 + x) + (0.25 + x + x^2) + (0.125 + 0.75x + 1.5x^2 + x^3) + \dots$$

With  $f = 2 + x^2$ ?

## Functions of formal power series

In general, if  $F$  is a function and  $f$  is a power series, we define

$$F(f) = F(f_0 + x) \circ (f_1 x + f_2 x^2 + \dots)$$

where

$$F(c + x) = F(c) + F'(c)x + \frac{1}{2}F''(c)x^2 + \dots$$

and  $\circ$  denotes **formal** composition of two power series.

Example:

$$\frac{1}{1-f} = \left( \frac{1}{1-f_0} + \frac{x}{(1-f_0)^2} + \frac{x^2}{(1-f_0)^3} \dots \right) \circ (f_1 x + f_2 x^3 + \dots)$$

# Fast power series arithmetic

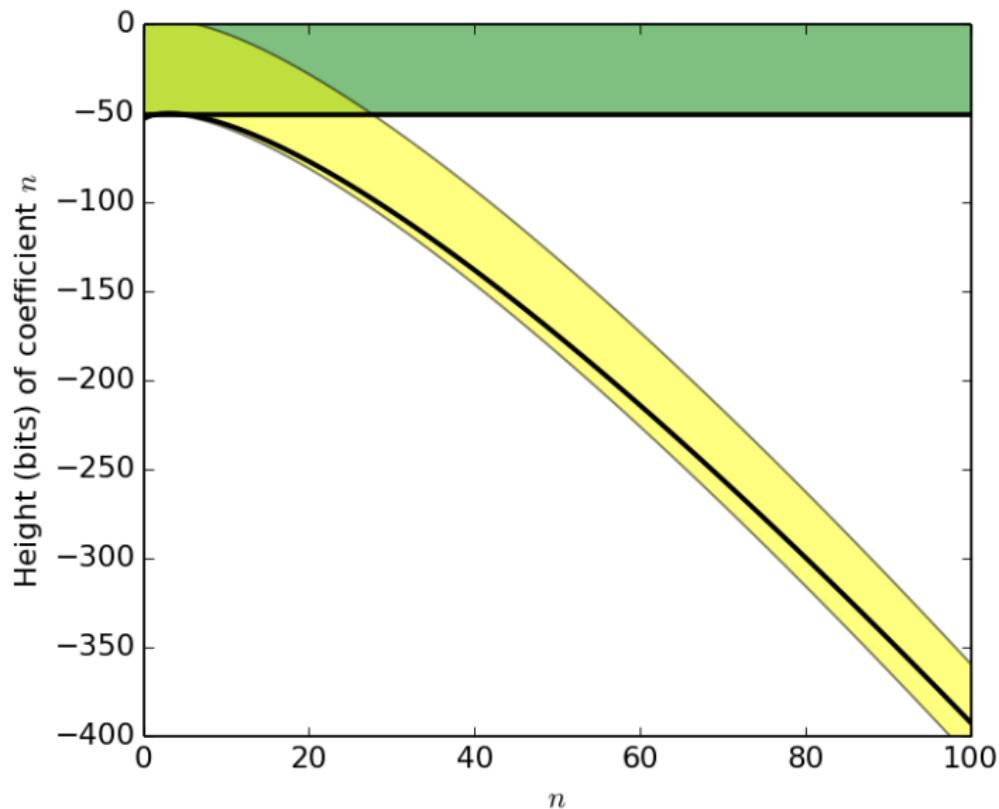
For large  $n$ , **reduce everything to multiplication**

- ▶ Classical polynomial multiplication:  $M(n) = O(n^2)$  arithmetic operations,  $O^\sim(n^2 p)$  bit operations
- ▶ FFT-based polynomial multiplication:  $M(n) = O^\sim(n)$  arithmetic operations,  $O^\sim(np)$  bit operations

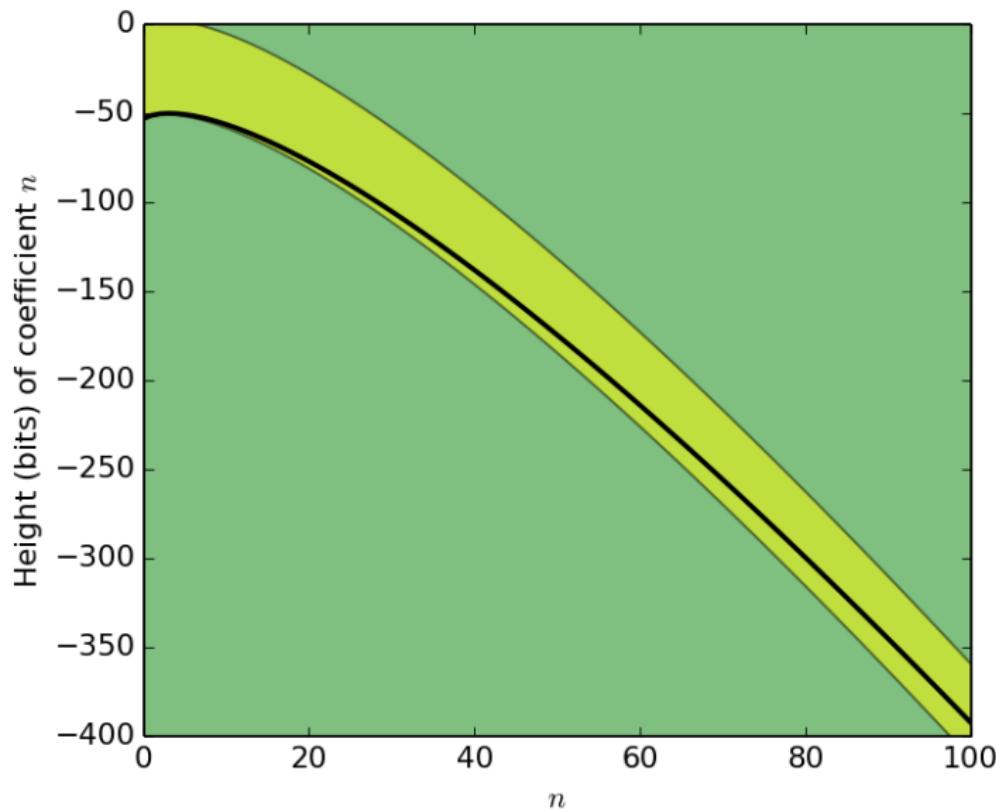
**Caveat:** When working with numerical data, we often need precision  $p \sim n$ , so effective bit complexity with FFT is often  $O^\sim(n^2)$ .

Feasible to work with  $n \approx 10^5$  ( $\approx 10^{10}$  bits of data).

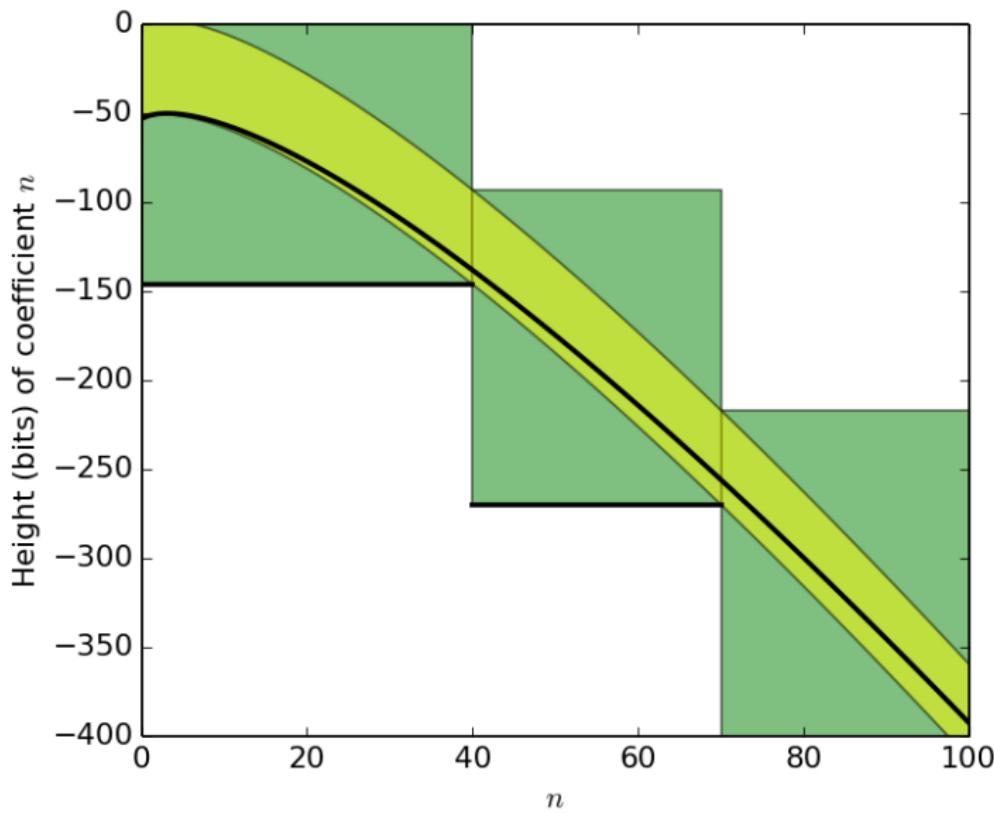
## Bad: truncate



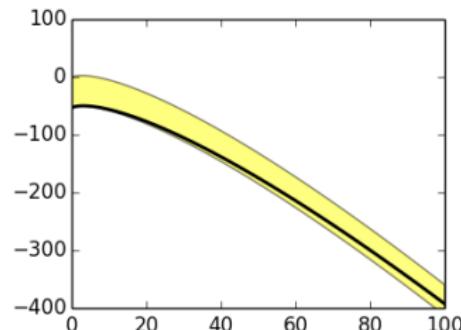
## Bad: cover everything



## Better: split into blocks

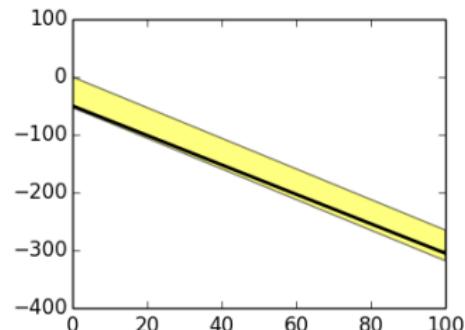


## Improving efficiency: scaling

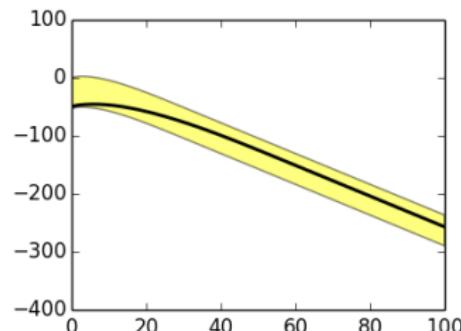


$$F(x)$$

$\times$

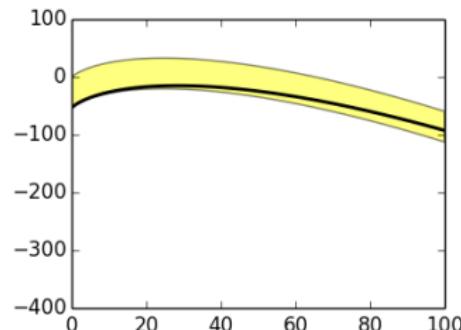


$$G(x)$$

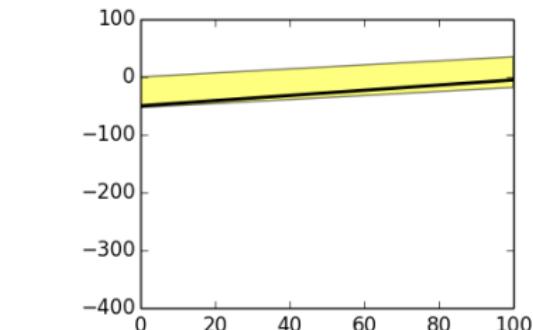


$$= H(x)$$

# Improving efficiency: scaling

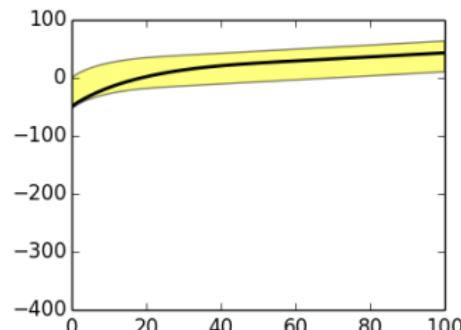


$$F(2^c x)$$



$$G(2^c x)$$

$\times$



$$= H(2^c x)$$

## Fast operations on power series

Many operations on power series of length  $n$  can all be done in  $O(M(n))$  arithmetic operations:

- ▶ Multiplication
- ▶ Division
- ▶ Square root
- ▶ Elementary functions ( $\exp$ ,  $\log$ ,  $\sin$ , ...)
- ▶ Solution of some differential equations

## Formal Newton iteration

To solve  $F(g) = 0$ , apply the Newton step

$$g_{\text{new}} = g_{\text{old}} - \frac{F(g_{\text{old}})}{F'(g_{\text{old}})}$$

If  $F(g_{\text{old}}) = O(x^n)$ , then  $F(g_{\text{new}}) = O(x^{2n})$

**Example:** to compute the reciprocal  $1/f$  of a power series, solve

$$F(g) = \frac{1}{g} - f$$

which corresponds to the Newton update

$$g_{\text{new}} = 2g_{\text{old}} - fg_{\text{old}}^2.$$

## Fast mass computation of Bernoulli numbers

Using the generating function:

$$\frac{x}{e^x - 1} = \frac{1}{1 + \frac{1}{2}x + \frac{1}{6}x^2 + \dots} = \sum_{k=0}^{\infty} \frac{B_k}{k!} x^k$$

Cost to compute  $n$  Bernoulli numbers exactly (using  $\mathbb{Q}[[x]]$  arithmetic):

- ▶  $O(M(n)) = O^{\sim}(n)$  arithmetic operations
- ▶  $O^{\sim}(n^2)$  bit operations

## A slower (but faster) algorithm for Bernoulli numbers

How to generate  $B_0, B_1, \dots, B_{10\,000}$  in two seconds!

$$B_{2n} = (-1)^{n+1} \frac{2(2n)!}{(2\pi)^{2n}} \zeta(2n)$$

$$\zeta(2) = 1 + \frac{1}{2^2} + \frac{1}{3^2} + \frac{1}{4^2} + \dots$$

$$\zeta(4) = 1 + \frac{1}{2^4} + \frac{1}{3^4} + \frac{1}{4^4} + \dots$$

$$\zeta(6) = 1 + \frac{1}{2^6} + \frac{1}{3^6} + \frac{1}{4^6} + \dots$$

Recycle the powers in each column (multiplication by  $2^2, 3^2, 4^2, \dots$ )

Complexity is  $O^\sim(n^3)$ , but in practice an order of magnitude faster than the  $O^\sim(n^2)$  power series algorithm. I first saw this algorithm in a blog by Remco Bloemen <http://2pi.com/09/11/even-faster-zeta-calculation>

## Elementary functions of power series

Functional equation + formal composition

$$\exp(f) = \exp(f_0) (\exp(x) \circ (f_1 x + f_2 x^2 + \dots))$$

$$\log(f) = \log(f_0) + \log(1+x) \circ \left( \frac{xf_1 + x^2 f_2 + \dots}{f_0} \right)$$

Fast formal composition ( $O(M(n))$  arithmetic operations):

$$\log(1+f) = \int \frac{f'}{1+f}$$

$$\text{atan}(f) = \int \frac{f'}{1+f^2}$$

Newton iteration gives  $\log \rightarrow \exp$ ,  $\text{atan} \rightarrow \tan$ , etc.

# Fast composition of arbitrary formal power series

Horner's rule

$$f(g(x)) = f_0 + g(f_1 + g(f_2 + \cdots + g(f_{n-2} + f_{n-1}x) \cdots ))$$

Arithmetic complexity:  $O(nM(n))$

Brent-Kung 2.1: baby-step, giant-step version of Horner's rule

Arithmetic complexity:  $O(n^{1/2}M(n) + n^{1/2}MM(n^{1/2}))$

Brent-Kung 2.2: divide-and-conquer Taylor expansion

Arithmetic complexity:  $O((n \log n)^{1/2}M(n))$

## Brent-Kung algorithm 2.1

Example with  $n = 9$ ,  $m = \lceil \sqrt{n} \rceil = 3$

Computing  $f(g)$  where  $f = f_0 + f_1x + \dots + f_8x^8$

$$(f_0 + f_1g + f_2g^2) + (f_3 + f_4g + f_5g^2)g^3 + (f_6 + f_7g + f_8g^2)g^6$$

$(m \times m) \times (m \times m^2)$  matrix multiplication:

$$\begin{pmatrix} f_0 & f_1 & f_2 \\ f_3 & f_4 & f_5 \\ f_6 & f_7 & f_8 \end{pmatrix} \times \begin{pmatrix} 1 \\ g \\ g^2 \end{pmatrix}$$

Finally apply Horner's rule to block polynomials.

## General functions of power series

If  $f, g$  are holomorphic with

$$f(z) \approx g(z)$$

for all  $z$  inside some disk, then also

$$f^{(n)}(z) \approx g^{(n)}(z)$$

$$\zeta(s) \approx \sum_{k=1}^N \frac{1}{k^s} \quad \Rightarrow \quad \zeta(s+x) \approx \sum_{k=1}^N \frac{1}{k^{s+x}}$$

where  $\zeta(s+x) = \zeta(s) + \zeta'(s)x + \frac{1}{2}\zeta''(s)x\dots$

To compute derivatives of  $\zeta(s)$ , take your favorite approximate formula and evaluate it at  $s+x \in \mathbb{C}[[x]]$  instead of  $s$ .

## Recall: the Euler-Maclaurin formula

$$\sum_{k=N}^U f(k) = I + T + R$$

$$I = \int_N^U f(t) dt$$

$$T = \frac{1}{2} (f(N) + f(U)) + \sum_{k=1}^M \frac{B_{2k}}{(2k)!} \left( f^{(2k-1)}(U) - f^{(2k-1)}(N) \right)$$

$$R = - \int_N^U \frac{\tilde{B}_{2M}(t)}{(2M)!} f^{(2M)}(t) dt$$

## Computing $\zeta(s, a)$ using Euler-Maclaurin

$$\zeta(s, a) = \underbrace{\sum_{k=0}^{N-1} f(k)}_S + \underbrace{\sum_{k=N}^{\infty} f(k)}_{I + T + R}, \quad f(k) = \frac{1}{(a+k)^s}$$

For derivatives, substitute  $s \rightarrow s + x \in \mathbb{C}[[x]]$ :

$$f(k) = \frac{1}{(a+k)^{s+x}} = \sum_{i=0}^{\infty} \frac{(-1)^i \log^i(a+k)}{i!(a+k)^s} x^i \in \mathbb{C}[[x]]$$

## The terms

$$f(1) = \left[ \frac{1}{(a+1)^s} \right] + \left[ -\frac{\log(a+1)}{(a+1)^s} \right] x + \left[ \frac{\log^2(a+1)}{2(a+1)^s} \right] x^2 + \dots$$

$$f(2) = \left[ \frac{1}{(a+2)^s} \right] + \left[ -\frac{\log(a+2)}{(a+2)^s} \right] x + \left[ \frac{\log^2(a+2)}{2(a+2)^s} \right] x^2 + \dots$$

$$f(3) = \left[ \frac{1}{(a+3)^s} \right] + \left[ -\frac{\log(a+3)}{(a+3)^s} \right] x + \left[ \frac{\log^2(a+3)}{2(a+3)^s} \right] x^2 + \dots$$

## Parts to evaluate

$$S = \sum_{k=0}^{N-1} \frac{1}{(a+k)^{s+x}}$$

$$I = \int_N^\infty \frac{1}{(a+t)^{s+x}} dt = \frac{(a+N)^{1-(s+x)}}{(s+x)-1}$$

$$T = \frac{1}{(a+N)^{s+x}} \left( \frac{1}{2} + \sum_{k=1}^M \frac{B_{2k}}{(2k)!} \frac{(s+x)_{2k-1}}{(a+N)^{2k-1}} \right)$$

$$R = - \int_N^\infty \frac{\tilde{B}_{2M}(t)}{(2M)!} \frac{(s+x)_{2M}}{(a+t)^{(s+x)+2M}} dt \quad (\text{bound})$$

## Bounding the remainder

Define  $|f| = |f_0| + |f_1|x + |f_2|x^2 + \dots$

Then  $|f + g| \leq |f| + |g|$  and  $|fg| \leq |f||g|$  (coefficient-wise)

$$\begin{aligned}|R| &= \left| \int_N^\infty \frac{\tilde{B}_{2M}(t)}{(2M)!} \frac{(s+x)_{2M}}{(a+t)^{s+x+2M}} dt \right| \\&\leq \int_N^\infty \left| \frac{\tilde{B}_{2M}(t)}{(2M)!} \frac{(s+x)_{2M}}{(a+t)^{s+x+2M}} \right| dt \\&\leq \frac{4|(s+x)_{2M}|}{(2\pi)^{2M}} \int_N^\infty \left| \frac{dt}{(a+t)^{s+x+2M}} \right| \in \mathbb{R}[[x]]\end{aligned}$$

$$\int_N^\infty \left| \frac{dt}{(a+t)^{s+x+2M}} \right| = \sum_{k=0}^{\infty} \left( \int_N^\infty \frac{dt}{k!} \left| \frac{\log(a+t)^k}{(a+t)^{s+2M}} \right| \right) x^k$$

## A sequence of integrals

For  $k \in \mathbb{N}$ ,  $A > 0$ ,  $B > 1$ ,  $C \geq 0$ ,

$$\begin{aligned} J_k(A, B, C) &\equiv \int_A^{\infty} t^{-B} (C + \log t)^k dt \\ &= \frac{L_k}{(B - 1)^{k+1} A^{B-1}} \end{aligned}$$

where

$$L_0 = 1, \quad L_k = kL_{k-1} + D^k$$

$$D = (B - 1)(C + \log A)$$

## Error bound

Given complex numbers  $s = \sigma + \tau i$ ,  $a = \alpha + \beta i$  and positive integers  $N, M$  such that  $\alpha + N > 1$  and  $\sigma + 2M > 1$ , the error term in the Euler-Maclaurin summation formula applied to  $\zeta(s + x, a) \in \mathbb{C}[[x]]$  satisfies

$$|R(s + x)| \leq \frac{4 |(s + x)_{2M}|}{(2\pi)^{2M}} \left| \sum_{k=0}^{\infty} R_k x^k \right| \in \mathbb{R}[[x]]$$

where  $R_k \leq (K/k!) J_k(N + \alpha, \sigma + 2M, C)$ , with

$$C = \frac{1}{2} \log \left( 1 + \frac{\beta^2}{(\alpha + N)^2} \right) + \operatorname{atan} \left( \frac{|\beta|}{\alpha + N} \right)$$

and

$$K = \exp \left( \max \left( 0, \tau \operatorname{atan} \left( \frac{\beta}{\alpha + N} \right) \right) \right).$$

## Evaluation steps

To evaluate  $\zeta(s + x, a)$  with an error of  $2^{-p}$ :

1. Choose  $N, M = O(p)$ , bound the error term  $R$
2. Compute the power sum  $S$
3. Compute the integral  $I$
4. Compute the Bernoulli numbers
5. Compute the tail  $T$

## Some computational results

## Stieltjes constants

The **(generalized) Stieltjes constants** are the coefficients  $\gamma_n(a)$  in the Laurent series

$$\zeta(s, a) = \frac{1}{s - 1} + \sum_{n=0}^{\infty} \frac{(-1)^n}{n!} \gamma_n(a) (s - 1)^n.$$

The **(classical) Stieltjes constants**  $\gamma_n = \gamma_n(1)$  have numerical values

$$\gamma_0 \approx +0.577216 \quad \gamma_{10} \approx +0.000205$$

$$\gamma_1 \approx -0.072816 \quad \gamma_{100} \approx -4.25340 \times 10^{17}$$

$$\gamma_2 \approx -0.009690 \quad \gamma_{1000} \approx -1.57095 \times 10^{486}$$

## Asymptotics of Stieltjes constants

One of the best available bounds for  $\gamma_n$  is [Matsuoka, 1985]:

$$|\gamma_n| < 0.0001e^{n \log \log n}, \quad n \geq 10$$

But this is not very accurate.

Actual value:  $\gamma_{1000} \approx -1.57095 \times 10^{486}$

Matsuoka:  $|\gamma_{1000}| < 2.17242 \times 10^{835}$

## Knessl-Coffey approximation

[Knessl and Coffey, *An effective asymptotic formula for the Stieltjes constants*, 2011]

$$\gamma_n \sim \frac{B}{\sqrt{n}} e^{nA} \cos(an + b)$$

$$A = \frac{1}{2} \log(u^2 + v^2) - \frac{u}{u^2 + v^2}, \quad B = \frac{2\sqrt{2\pi}\sqrt{u^2 + v^2}}{[(u+1)^2 + v^2]^{1/4}}$$

$$a = \tan^{-1} \left( \frac{v}{u} \right) + \frac{v}{u^2 + v^2}, \quad b = \tan^{-1} \left( \frac{v}{u} \right) - \frac{1}{2} \left( \frac{v}{u+1} \right)$$

where  $u = v \tan v$ , and  $v$  is the unique solution of  
 $2\pi \exp(v \tan v) = (n/v) \cos(v)$ ,  $0 < v < \pi/2$ . Similar formula for  $\gamma_n(a)$ ,  
 $a \neq 1$ .

- ▶ Predicts sign oscillations (correct except for  $n = 137$ ?)
- ▶ Accurate even for small  $n$
- ▶ No explicit error bound

# Computations of Stieltjes constants

[Kreminski, *Newton-Cotes integration for approximating Stieltjes (generalized Euler) constants*, 2003]:

- ▶ Values up to  $n = 8300$  with 200 significant digits
- ▶ Heuristic error estimates

Later heuristic computations up to about  $n = 35\,000$

[FJ, 2014]:

- ▶ All  $\gamma_n$  up to  $n = 100\,000$  with more than 10 000 digits each
- ▶ Rigorous error bounds
- ▶ To replicate, simply call the Hurwitz zeta function of a power series in Arb (also for  $\gamma_n(a)$ )

Data available from:

[http://fredrikj.net/math/hurwitz\\_zeta.html](http://fredrikj.net/math/hurwitz_zeta.html)

## Numerical values

Computed value of  $\gamma_{100000}$ :

$$1.99192730631254109565 \dots \times 10^{83432}$$

Knessl-Coffey approximation:

$$1.9919\textcolor{red}{333} \times 10^{83432}$$

Matsuoka bound:  $3.71 \times 10^{106114}$

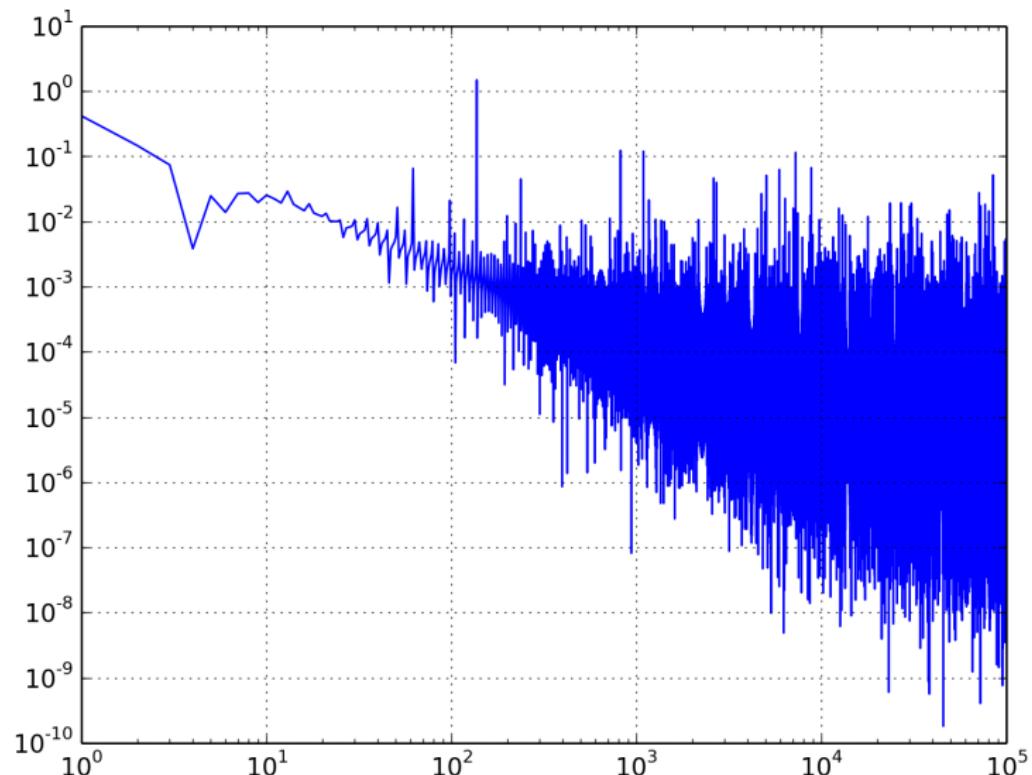
Computed value of  $\lambda_{50000}(1 + i)$ :

$$(1.032502087431 \dots - 1.441962552840 \dots i) \times 10^{39732}$$

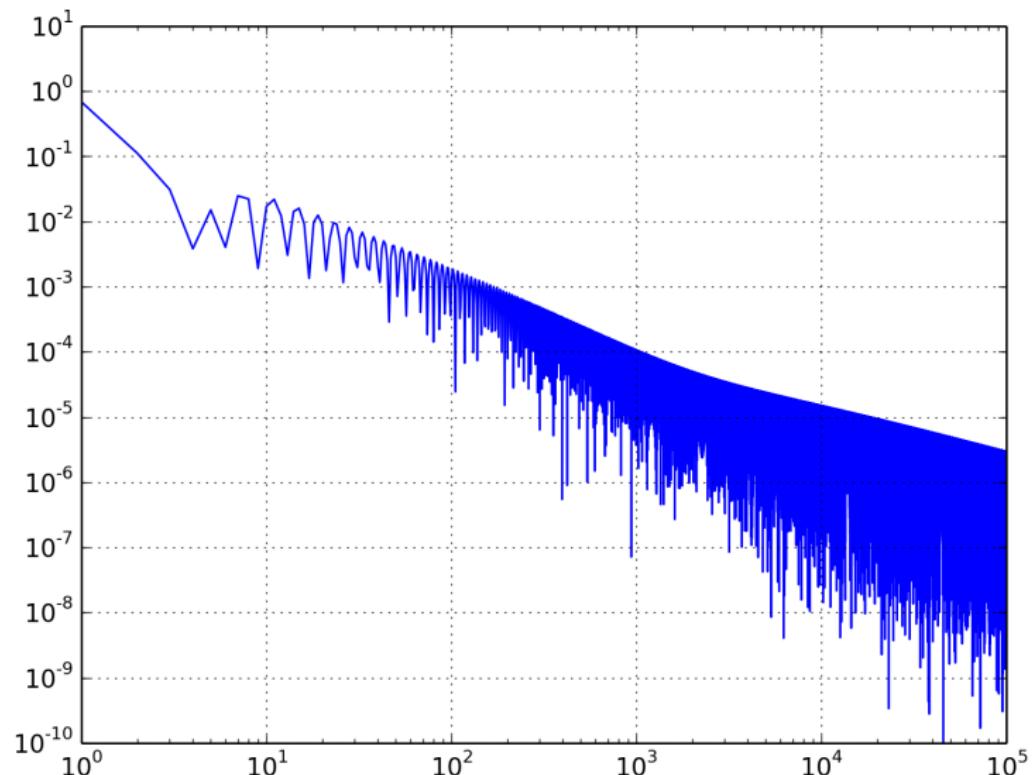
Knessl-Coffey approximation:

$$(\textcolor{green}{1.0324943} - \textcolor{green}{1.4419586}i) \times 10^{39732}$$

## Relative error of Knessl-Coffey formula



# Relative error of Knessl-Coffey formula



# The Keiper-Li coefficients

Define  $\{\lambda_n\}_{n=1}^{\infty}$  by

$$\log \xi \left( \frac{1}{1-x} \right) = \log \xi \left( \frac{x}{x-1} \right) = -\log 2 + \sum_{n=1}^{\infty} \lambda_n x^n$$

where  $\xi(s) = \frac{1}{2}s(s-1)\pi^{-s/2}\Gamma(s/2)\zeta(s)$ .

Keiper (1992): Riemann hypothesis  $\Rightarrow \forall n : \lambda_n > 0$

Li (1997): Riemann hypothesis  $\Leftarrow \forall n : \lambda_n > 0$

Keiper conjectured  $2\lambda_n \approx (\log n - \log(2\pi) + \gamma - 1)$

# Evaluating the Keiper-Li coefficients

Evaluate

$$\log \xi(s) = \log(-\zeta(s)) + \log \Gamma(1 + s/2) + \log(1 - s) - s \log(\pi)/2$$

at  $s = x \in \mathbb{R}[[x]]$

Ingredients:

1. The series expansion  $\zeta(s + x)$  at  $s = 0$
2. The logarithm of a power series:  $\log f(x) = \int f'(x)/f(x)dx$
3. The series  $\log \Gamma(1 + x)$ , essentially  $\gamma, \zeta(2), \zeta(3), \zeta(4), \dots$
4. Right-composing by  $x/(x - 1)$

A working precision of  $\approx n$  bits is needed to get an accurate value for  $\lambda_n$ .

## Fast composition

The *binomial transform* of  $f = \sum_{k=0}^{\infty} a_k x^k$  is

$$T[f] = \frac{1}{1-x} f\left(\frac{x}{x-1}\right) = \sum_{n=0}^{\infty} \left( \sum_{k=0}^n (-1)^k \binom{n}{k} a_k \right) x^n$$

and the *Borel transform* is

$$B[f] = \sum_{k=0}^{\infty} \frac{a_k}{k!} x^k.$$

$T[f(x)] = B^{-1}[e^x B[f(-x)]]$ , so we get  $f\left(\frac{x}{x-1}\right)$  by a single power series multiplication!

## Values of Keiper-Li coefficients

I have computed rigorous values of all  $\lambda_n$  up to  $n = 100\,000$  (using Arb, and 110 000 bits of working precision). In particular,

$$\lambda_{100000} = 4.62580782406902231409416038\dots$$

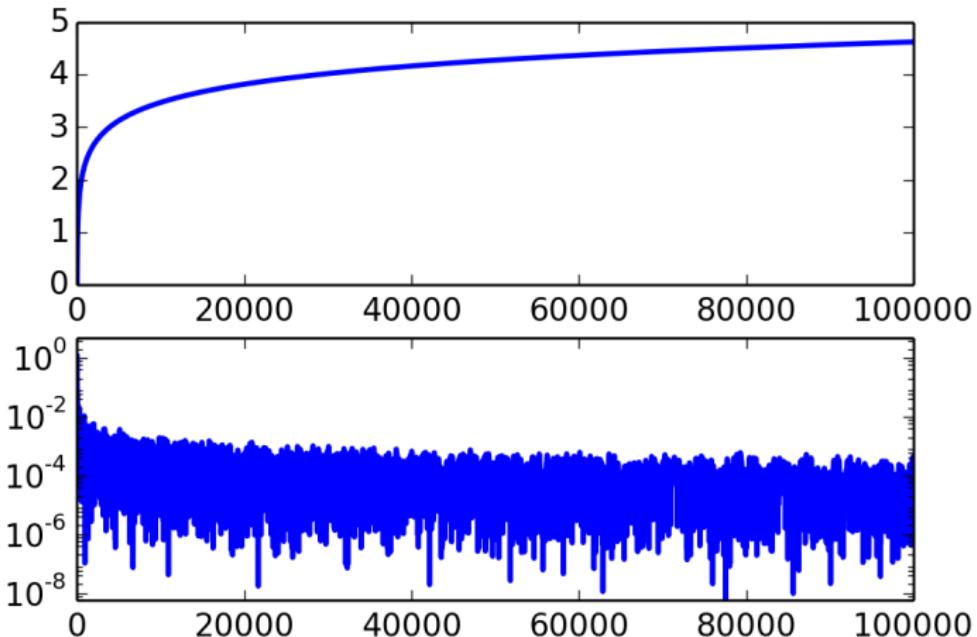
plus about 2900 more accurate digits.

Keiper's approximation suggests  $\lambda_{100000} \approx 4.626132$ .

See examples/keiper\_li.c in Arb and

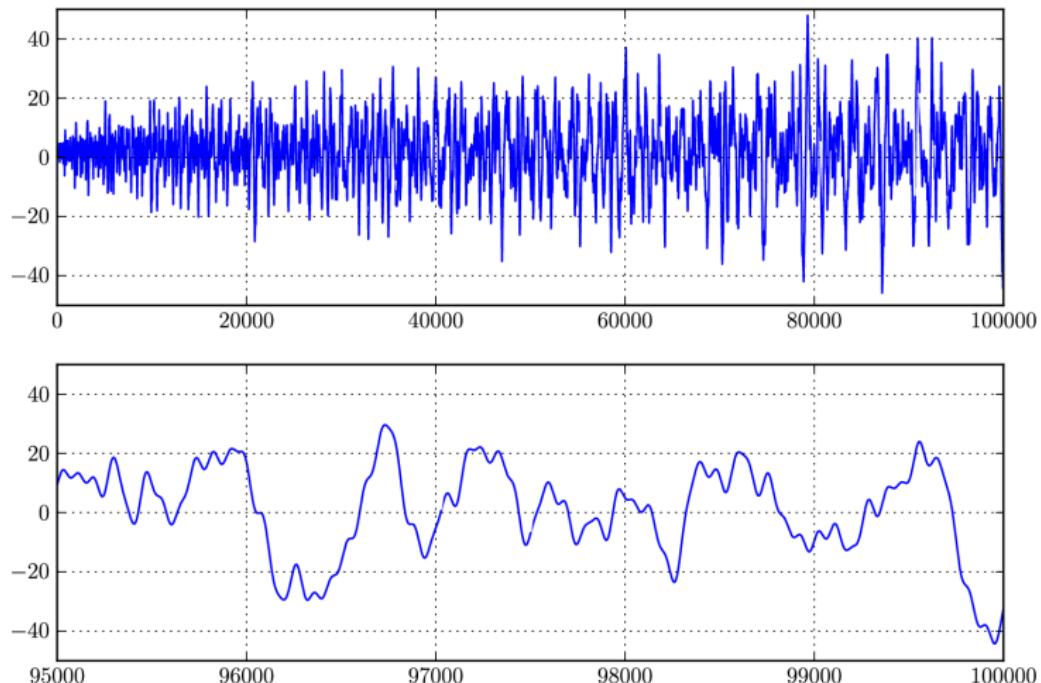
[http://fredrikj.net/math/hurwitz\\_zeta.html](http://fredrikj.net/math/hurwitz_zeta.html) for data

## Comparison with approximation formula



Plot of  $\lambda_n$  and  $\lambda_n - (\log n - \log(2\pi) + \gamma - 1)/2$

## Comparison with approximation formula



Plot of  $n(\lambda_n - (\log n - \log(2\pi) + \gamma - 1)/2)$ .

## Time to compute Keiper-Li coefficients

(In seconds)

	$n = 1000$	$n = 10000$	$n = 100000$
Error bound $R$	0.017	1.0	97
Power sum $S$ (CPU time)	0.048 (0.65)	47 (693)	65402 (1042210)
Bernoulli numbers	0.0020	0.19	59
Tail $T$	0.058	11	1972
Logarithm of power series	0.047	8.5	1126
$\log \Gamma(1 + x)$	0.019	3.0	1610
Composition by $x/(x - 1)$	0.022	4.1	593
Total wall time	0.23	84	71051
RAM (MiB)	8	730	48700

Bit complexity is  $O^\sim(n^2)$  except for the power sum which is  $O^\sim(n^3)$   
( $O^\sim(n^2)$  is possible in theory, but not implemented)