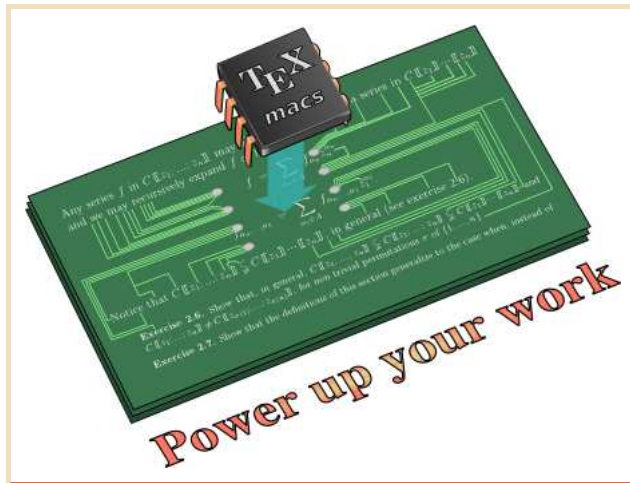


The art of guessing

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<http://www.TEXMACS.org>



Guess what



0.142857142857142857142857142857142857...

0.142857142857142857142857142857142857...

$$\frac{1}{7}$$

Applications:

- More comprehensive output.
- New unexpected insights.
- Guide decisions in algorithms.



Efficient algorithm



$$x_0 = a_0 + \frac{1}{x_1} \quad a_0 = \lfloor x_0 \rfloor$$

$$x_0 = a_0 + \frac{1}{a_1 + \frac{1}{x_2}} \quad a_1 = \lfloor x_1 \rfloor$$

$$x_0 = a_0 + \frac{1}{a_1 + \frac{1}{\ddots + \frac{1}{a_{l-2} + \frac{1}{a_{l-1} + x_l}}}}$$

$$x_l = \frac{p_l x_0 + q_l}{r_l x_0 + s_l} \quad p_l, q_l, r_l, s_l \in \mathbb{Z}$$

$$x_{k+l} = \frac{p_{k,l} x_l + q_{k,l}}{r_{k,l} x_l + s_{k,l}} \quad p_{k,l}, q_{k,l}, r_{k,l}, s_{k,l} \in \mathbb{Z}$$

$$\begin{pmatrix} p_{k,l+m} & q_{k,l+m} \\ r_{k,l+m} & s_{k,l+m} \end{pmatrix} = \begin{pmatrix} p_{k+l,m} & q_{k+l,m} \\ r_{k+l,m} & s_{k+l,m} \end{pmatrix} \begin{pmatrix} p_{k,l} & q_{k,l} \\ r_{k,l} & s_{k,l} \end{pmatrix}$$

$$\begin{aligned} F(p) &= 2F(p/2) + O(\log p) \\ &= O(\log p \log p) \\ &= O(p \log^2 p \log \log p) \end{aligned}$$



Guessing $\mathbb{Z}$ -linear dependencies



Question: given $z_1, \dots, z_n \in \mathbb{R}$, find $\lambda_1, \dots, \lambda_n \in \mathbb{Z}$ with

$$\lambda_1 z_1 + \dots + \lambda_n z_n = 0$$

Example:

- $z_1 = 1, \dots, z_n = \alpha^{n-1}$ for $\alpha \in \mathbb{R}$.
- $z_1 = \text{Li}_{3,3}, z_2 = \text{Li}_6, z_3 = \text{Li}_{4,2}, z_4 = \text{Li}_{2,2,2}$.

Lattice reduction:

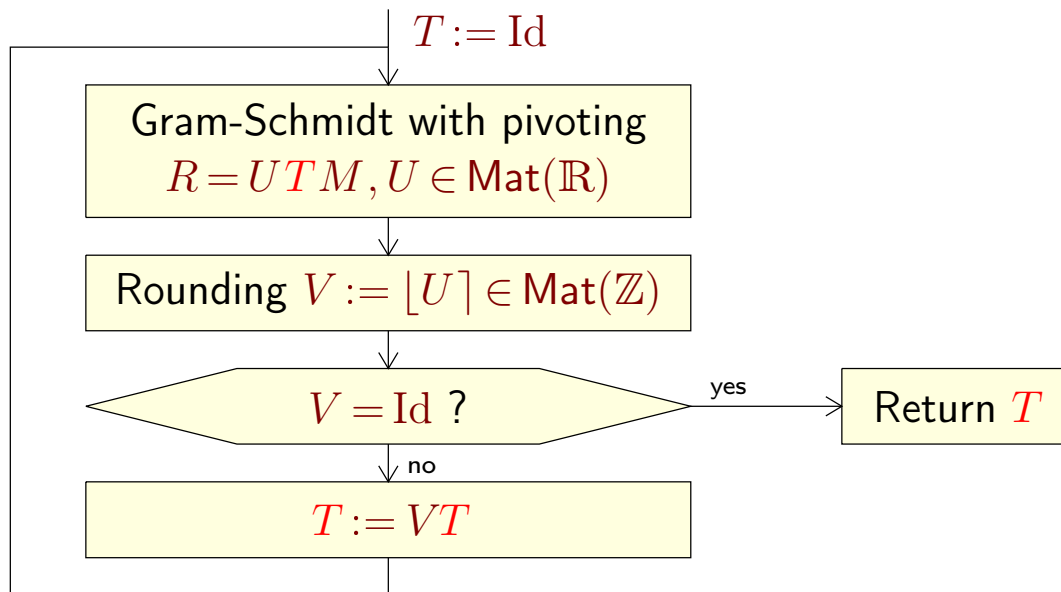
$$M = \begin{pmatrix} z_1 & \varepsilon & 0 & 0 \\ \vdots & 0 & \ddots & 0 \\ z_n & 0 & 0 & \varepsilon \end{pmatrix}$$



Block lattice reduction (work in progress)



The main loop



Dichotomy over precision

1. For low precisions, compute $T \in \text{Mat}(\mathbb{Z})$ as above.

2. For large precisions p , do

$$T_1 := \text{Reduce}(M \parallel 2^{-p/2} |M| \text{Id}) \quad (\text{at precision } p/2)$$

$$T_2 := \text{Reduce}(T_1 M) \quad (\text{at precision } p/2)$$

and return $T_2 T_1$.



Power series analogues



Guessing rationality

Problem: given $f \in \mathbb{K}[[z]]$, guess $P, Q \in \mathbb{K}[z]$ with $f = \frac{P}{Q}$

Solution: Padé approximation

Polynomial dependencies

Problem: given $f_1, \dots, f_n \in \mathbb{K}[[z]]$, guess $P_1, \dots, P_n \in \mathbb{K}[z]$ with

$$P_1 f_1 + \dots + P_n f_n = 0.$$

Solution: Padé-Hermite (Beckermann-Labahn, Derksen, Gfun)

Applications

- $(f_n)_{n \in \mathbb{N}} = 1, 1, 2, 3, 5, 8, 13, 21, 34, 55, 89, \dots$
- $f_1 = \varphi, f_2 = \varphi', \dots, f_n = \varphi^{(n-1)}$, with $\varphi \in \mathbb{K}[[z]]$



Guessing singular dependencies



Example: number s_n of alcohols of the form $C_n H_{2n+1} O H$

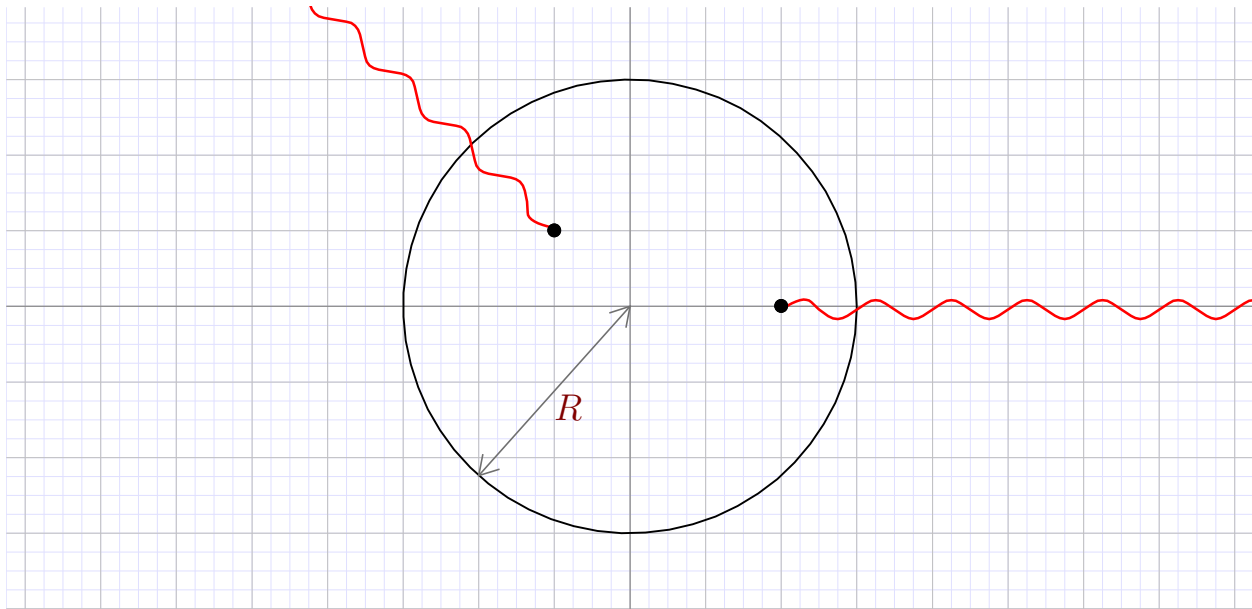
$$s(z) = 1 + z \frac{s(z)^3 + 2 s(z^3)}{3}.$$

Dominant **algebraic** singularity at $r = 0.304218409\dots$. Setting

$$\begin{aligned}\varphi(z) &= s(r+z) \\ \psi(z) &= 2 s((r+z)^3) + 3,\end{aligned}$$

we have

$$(r+z) \varphi(z)^3 - 3 \varphi(z) + \psi(z) = 0.$$



Problem: given $f_1, \dots, f_d \in \mathbb{C}[[z]]$ and R , guess relations

$$g_1 f_1 + \dots + g_d f_d = h \quad (g_1, \dots, g_d, h \in \mathbb{A}_R; \text{ i.e. analytic on } \bar{\mathcal{D}}_R)$$

Problem: given $f_1, \dots, f_d \in \mathbb{C}[[z]]$ and $R = 1$, find a “minimal relation”

$$g_1 f_1 + \dots + g_d f_d = h$$

for the power series norm

$$\|\varphi\| = \sqrt{\varphi_0^2 + \varphi_1^2 + \dots}.$$



Truncated problem at order n



$$g_1 f_1 + \cdots + g_d f_d = h$$

$$G = \begin{pmatrix} g_{1,n-1} & \cdots & g_{d,n-1} & \cdots & \cdots & g_{1,0} & \cdots & g_{d,0} \end{pmatrix}$$

$$M = \begin{pmatrix} f_{1,0} & & & & & 1 & & & & & \\ \vdots & & & & & \ddots & & & & & \\ f_{d,0} & & & & & & 1 & & & & \\ f_{1,1} & f_{1,0} & & & & & & 1 & & & \\ \vdots & \vdots & & & & & & \ddots & & & \\ f_{d,1} & f_{d,0} & & & & & & & 1 & & \\ \vdots & \vdots & & & & & & & \ddots & & \\ \vdots & \vdots & & \ddots & \ddots & & & & & \ddots & \\ f_{1,n-1} & f_{1,n-2} & \cdots & \cdots & f_{1,0} & & & & & 1 & \ddots \\ \vdots & \vdots & & & \vdots & & & & & \ddots & \\ f_{d,n-1} & f_{d,n-2} & \cdots & \cdots & f_{d,0} & & & & & & 1 \end{pmatrix}.$$

$$G M = \begin{pmatrix} h_{n-1} & h_{n-2} & \cdots & \cdots & h_0 & g_{1,n-1} & \cdots & g_{d,n-1} & \cdots & g_{1,0} & \cdots & g_{d,0} \end{pmatrix}$$



Ultimate correctness



Definition. Given $n \in \mathbb{N} \cup \{+\infty\}$, let $\Phi_{;n}$ be the set of vectors

$$\varphi = (h_{n-1}, \dots, h_0, g_{1,n-1}, \dots, g_{d,n-1}, \dots, g_{1,0}, \dots, g_{d,0}),$$

$$\|\varphi\| < +\infty$$

We say that φ is ***i-normal*** if $g_{i,0} = 1$ and $g_{j,0} = 0$ for all $j > i$.

Theorem.

- 1) If $\Phi_{;\infty} \neq 0$, then $\Phi_{;\infty}$ contains a minimal normal relation.
- 2) Assume that $\varphi \in \Phi_{;\infty}$ is a minimal *i-normal* relation. For each $n \in \mathbb{N}$, let $f_{;n}$ be the truncation of f at order n and consider the corresponding minimal *i-normal* relation $\varphi_{;n} \in \Phi_{;n}$. Then the relations $\varphi_{;n}$ converge to φ .
- 3) If $\Phi_{;\infty} = 0$, then the $\|\varphi_{;n}\|$ are unbounded for $n \rightarrow \infty$.



Example I



$$f_1 = \frac{1}{1 - \lambda z} \quad (\lambda > 1)$$

$$g_{;4,1} = 1.0000 - 1.6000 z - 0.60000 z^2 \\ - 0.20000 z^3$$

$$g_{;16,1} = 1.0000 - 1.6180 z - 0.61803 z^2 - \dots \\ - 5.8340 \cdot 10^{-6} z^{14} - 1.9447 \cdot 10^{-6} z^{15}$$

$$g_{;64,1} = 1.0000 - 1.6180 z - 0.61803 z^2 - \dots \\ - 5.0484 \cdot 10^{-26} z^{62} - 1.6828 \cdot 10^{-26} z^{63}$$

$$g_{;256,1} = 1.0000 - 1.6180 z - 0.61803 z^2 - \dots \\ - 2.8307 \cdot 10^{-106} z^{254} - 9.4357 \cdot 10^{-107} z^{255}$$

Fast convergence, but **not** $g_1 = 1 - \lambda z$. In fact

$$g_1 = \frac{1 - \lambda z}{1 - \alpha z}$$

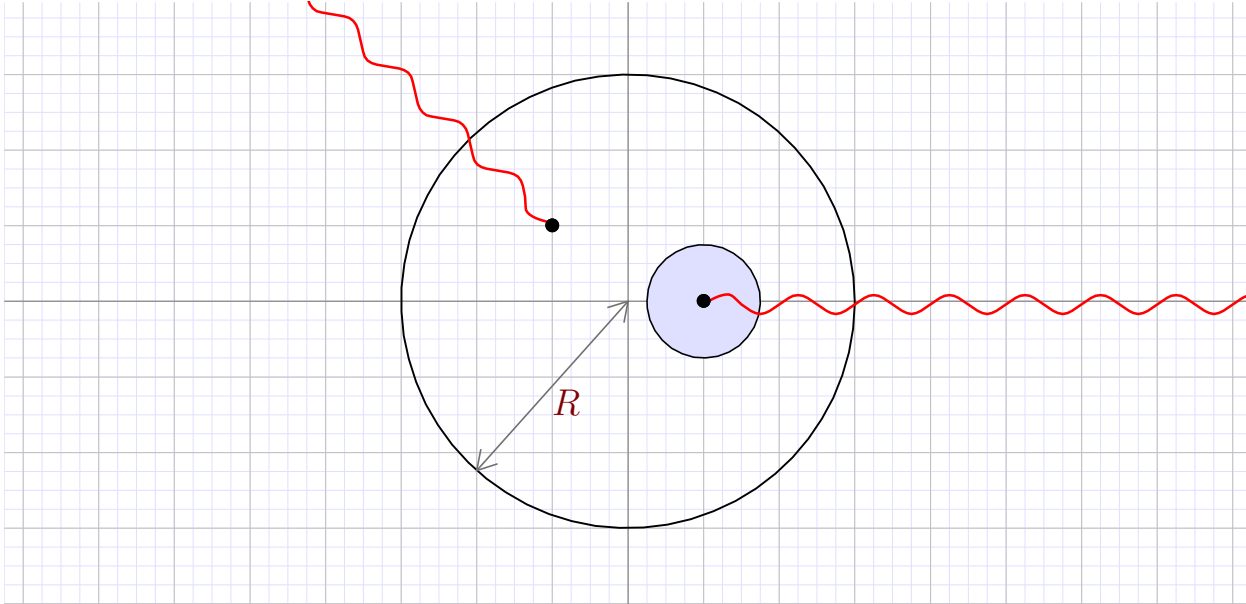
$$\alpha + \frac{1}{\alpha} = \lambda + \frac{2}{\lambda} \quad (\alpha < 1)$$



Example II



$$\begin{aligned} f_1 &= s(0.15(z + 0.25)) \\ f_2 &= f_1' \end{aligned}$$



$$\begin{aligned} g_{;32,1} &= 1.6836 - 0.29080z + 0.0013079z^2 + \dots \\ &\quad + 0.022645z^{30} - 0.0029284z^{31} \end{aligned}$$

$$\begin{aligned} g_{;32,2} &= 1.0000 - 2.4006z - 0.88670z^2 - \dots \\ &\quad + 0.0061880z^{30} - 9.9191 \cdot 10^{-4}z^{31} \end{aligned}$$

$$\begin{aligned} g_{;64,1} &= 1.7104 - 0.36445z + 0.23778z^2 - \dots \\ &\quad + 1.6968 \cdot 10^{-4}z^{62} - 1.0104 \cdot 10^{-5}z^{63} \end{aligned}$$

$$\begin{aligned} g_{;64,2} &= 1.0000 - 2.3257z - 1.2027z^2 - \dots \\ &\quad + 5.2359 \cdot 10^{-5}z^{62} - 3.4225 \cdot 10^{-6}z^{63} \end{aligned}$$

$$\begin{aligned} g_{;128,1} &= 1.7105 - 0.36215z + 0.23696z^2 - 0.052535z^3 + 0.033518z^4 - \dots \\ &\quad + 4.6860 \cdot 10^{-9}z^{126} - 1.3437 \cdot 10^{-10}z^{127} \end{aligned}$$

$$\begin{aligned} g_{;128,2} &= 1.0000 - 2.3235z - 1.2138z^2 - 0.0067226z^3 - 0.080434z^4 + \dots \\ &\quad + 1.5192 \cdot 10^{-9}z^{126} - 4.5516 \cdot 10^{-11}z^{127} \end{aligned}$$

Slow convergence: $\|\varphi_{;32}\| = 4.3276$, $\|\varphi_{;64}\| = 4.3845$, $\|\varphi_{;128}\| = 4.3863$.



Counter-Example III



$$f_1 = \psi_{i,\lambda}$$

$$\psi_{1,\lambda} = \log(1 - \lambda z)$$

$$\psi_{2,\lambda} = \sqrt{1 - \lambda z}$$

$$\psi_{3,\lambda} = e^{\frac{\lambda z}{1 - \lambda z}}$$

$$\psi_{4,\lambda} = \text{random}(\lambda z)$$

	32	64	128	256
$\psi_{1,\sqrt{2}}$	2.5897	4.2958	$1.1107 \cdot 10^1$	$7.5308 \cdot 10^1$
$\psi_{1,2}$	$1.2324 \cdot 10^1$	$7.7306 \cdot 10^1$	$3.3839 \cdot 10^3$	$6.4461 \cdot 10^6$
$\psi_{1,4}$	$2.7074 \cdot 10^3$	$3.1503 \cdot 10^6$	$5.0806 \cdot 10^{12}$	$1.3215 \cdot 10^{25}$
$\psi_{1,8}$	$5.7101 \cdot 10^6$	$1.6078 \cdot 10^{13}$	$1.2829 \cdot 10^{26}$	$8.6064 \cdot 10^{51}$
$\psi_{1,16}$	$6.0964 \cdot 10^{10}$	$1.7814 \cdot 10^{21}$	$1.5612 \cdot 10^{42}$	$1.2279 \cdot 10^{84}$
$\psi_{1,32}$	$1.1152 \cdot 10^{15}$	$9.1674 \cdot 10^{29}$	$4.9020 \cdot 10^{59}$	$1.6662 \cdot 10^{119}$
$\psi_{2,\sqrt{2}}$	2.2960	3.7208	9.7394	$6.5890 \cdot 10^1$
$\psi_{2,2}$	9.3539	$6.1551 \cdot 10^1$	$2.6678 \cdot 10^3$	$5.0111 \cdot 10^6$
$\psi_{2,4}$	$1.7232 \cdot 10^3$	$2.1500 \cdot 10^6$	$3.4149 \cdot 10^{12}$	$8.7128 \cdot 10^{24}$
$\psi_{2,8}$	$3.5980 \cdot 10^6$	$1.0031 \cdot 10^{13}$	$7.9396 \cdot 10^{25}$	$5.2284 \cdot 10^{51}$
$\psi_{3,\sqrt{2}}$	6.2155	$1.3295 \cdot 10^1$	$4.1722 \cdot 10^1$	$4.2679 \cdot 10^2$
$\psi_{3,2}$	$4.0164 \cdot 10^1$	$3.9310 \cdot 10^2$	$1.4047 \cdot 10^4$	$1.1091 \cdot 10^7$
$\psi_{3,4}$	$6.9272 \cdot 10^3$	$6.8304 \cdot 10^6$	$4.7564 \cdot 10^{11}$	$5.1308 \cdot 10^{20}$
$\psi_{3,8}$	$3.6660 \cdot 10^6$	$9.4896 \cdot 10^{11}$	$4.2745 \cdot 10^{21}$	$1.1809 \cdot 10^{39}$
$\psi_{4,2}$	$8.0487 \cdot 10^4$	$3.5565 \cdot 10^9$	$2.1354 \cdot 10^{19}$	$4.8792 \cdot 10^{38}$
$\psi_{4,4}$	$5.0774 \cdot 10^9$	$1.1548 \cdot 10^{19}$	$2.9916 \cdot 10^{38}$	$1.2335 \cdot 10^{77}$
$\psi_{4,8}$	$3.2564 \cdot 10^{14}$	$4.3151 \cdot 10^{28}$	$5.1348 \cdot 10^{57}$	$2.9654 \cdot 10^{115}$

$$\log \|\varphi_n\| \approx \frac{n}{2} \log \lambda.$$

$$\log \|\varphi_n\| \approx \frac{n}{d+1} \log \lambda.$$



Only poles



Problem: given $f \in \mathbb{C}[[z]]$, guess $P \in \mathbb{C}[z]$ with

$$Pf \in \mathbb{A}_1.$$

Algorithm $\text{denom}(f, N, D)$

INPUT: the first $N > 2D$ coefficients of f and a degree bound D

OUTPUT: $\approx$ minimal monic polynomial P with $\deg P \leq D$, $Pf \in \mathbb{A}_1$, or **failed**

Step 1. [Initialize]

$$d := 0$$

Step 2. [Determine P]

$$\text{Solve } \begin{pmatrix} f_{N-2d+1} & \cdots & f_{N-d} \\ \vdots & & \vdots \\ f_{N-d} & \cdots & f_{N-1} \end{pmatrix} \begin{pmatrix} P_{d-1} \\ \vdots \\ P_0 \end{pmatrix} + \begin{pmatrix} f_{N-2d} \\ \vdots \\ f_{N-d-1} \end{pmatrix} = 0.$$

$$\text{Set } P := z^d + P_{d-1}z^{d-1} + \cdots + P_0$$

Step 3. [Terminate or loop]

If $Pf \in \mathbb{A}_1$ then return P

If $d = N$ then return **failed**

Set $d := d + 1$ and go to step 2

Theorem 1. *Exponential convergence in N .*



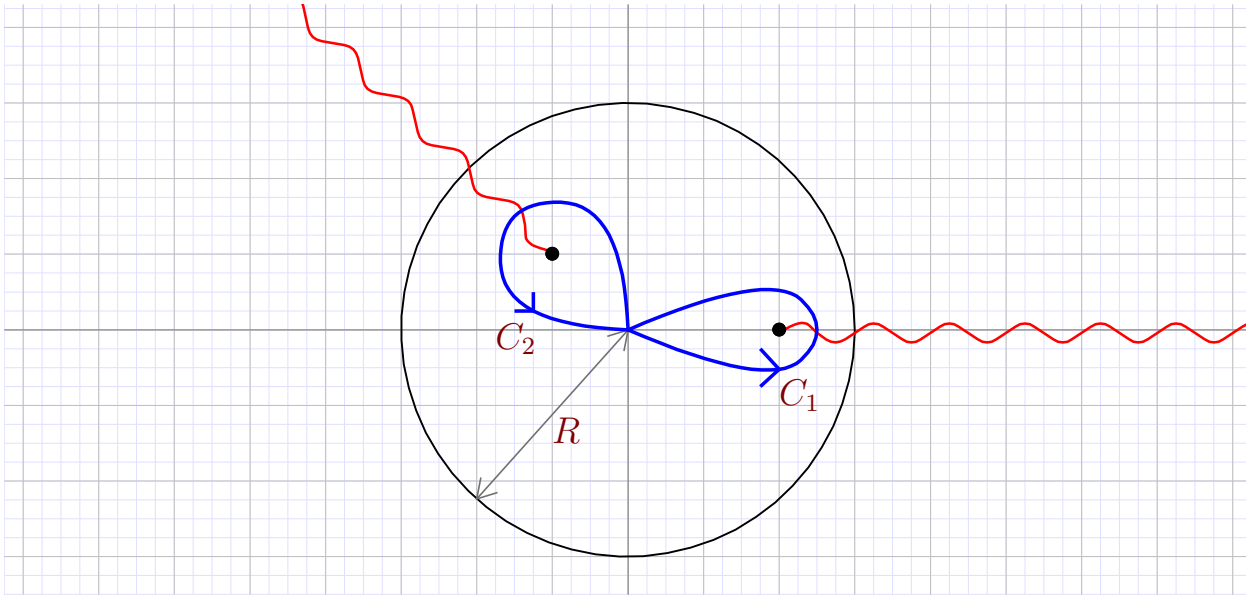
Only algebraic singularities



Problem: given $f \in \mathbb{C}[[z]]$, guess $P_d, \dots, P_0 \in \mathbb{C}[z]$ with

$$P_d f^d + \dots + P_0 \in \mathbb{A}_1. \quad (1)$$

Assume: algorithm for analytic continuation of f .



Algorithm alg_dep($f, (\sigma_1, \dots, \sigma_s), N, D, B$)

INPUT: analytic function f above $\bar{\mathcal{D}}_1 \setminus \{\sigma_1, \dots, \sigma_s\}$ and bounds N, D, B

OUTPUT: normalized dependency (1) with $d \leq B$, $\deg P_d \leq D$, or **failed**

Step 1. [Initialize]

Set $\Phi := \{f\}$

Step 2. [Saturate]

If $C_i \Phi \subseteq \Phi$ for all i , then go to step 3

If $\text{card}(\Phi) > N$ then return **failed**

$\Phi := C_1 \Phi \cup \dots \cup C_s \Phi$

Repeat step 2

Step 3. [Terminate]

Denote $\Phi := \{\varphi_1, \dots, \varphi_k\}$

Compute $Q := (F - \varphi_1) \dots (F - \varphi_k) = Q_k F^k + \dots + Q_0$

For each $i \in \{0, \dots, k\}$, compute $D_i := \text{denom}(Q_i, r, D)$

If $D_i = \text{failed}$ for some i , then return **failed**

Return $\text{lcm}(D_0, \dots, D_k) Q$



Only Fuchsian singularities



Problem: given $f \in \mathbb{C}[[z]]$, guess $L_r, \dots, L_0 \in \mathbb{C}[z]$ with

$$L_r f^{(r)} + \dots + L_0 \in \mathbb{A}_1. \quad (2)$$

Assume: algorithm for analytic continuation of f .

Algorithm fuch_dep($f, (\sigma_1, \dots, \sigma_s), N, D, B$)

INPUT: analytic function f above $\bar{\mathcal{D}}_1 \setminus \{\sigma_1, \dots, \sigma_s\}$ and bounds N, D, B

OUTPUT: normalized dependency (2) with $r \leq B, \deg L_r \leq D$, or **failed**

Step 1. [Initialize]

Set $\Phi := \{f\}$

Step 2. [Saturate]

If $\text{Vect}(C_i \Phi) \subseteq \text{Vect}(\Phi)$ for all i , then go to step 3

If $\text{card}(\Phi) > N$ then return **failed**

$\Phi := \Phi \cup \{C_i \varphi\}$ for i and $\varphi \in \Phi$ with $C_i \varphi \notin \text{Vect}(\Phi)$

Repeat step 2

Step 3. [Terminate]

Denote $\Phi := \{\varphi_1, \dots, \varphi_k\}$

Compute $K := \text{lcm}(\partial - \varphi_1^\dagger, \dots, \partial - \varphi_k^\dagger) = K_k \partial^k + \dots + K_0$

in the skew polynomial ring $\mathbb{C}((z))[\partial]$, where $\varphi^\dagger$ denotes φ'/φ

For each $i \in \{0, \dots, k\}$, compute $D_i := \text{denom}(K_i, r, D)$

If $D_i = \text{failed}$ for some i , then return **failed**

Return $\text{lcm}(D_0, \dots, D_k) K$

Remark. Only works for Fuchsian singularities, because of

$$f = e^{\frac{1}{z-\sigma}} + e^{\frac{-1}{z-\sigma}}.$$



Guessing asymptotic behaviour



Guessing Stirling's formula

```
Mmx] use "symbolix"; use "multimix"; use "jorix"
```

```
Mmx] include "jorix/extrapolate_extra.mmx"
```

```
Mmx] bit_precision := 256; significant_digits := 12;
```

```
Mmx] v == [ 1.0 * n! | n in 0..1000 ];
```

```
Mmx] guess_asymptotics (v, 7)
```

$$2.50662828099 e^{1.000000000000n \log(n) - 0.999999999999n + 0.499999999637 \log(n)} + \\ 0.208884786159 e^{1.000000000000n \log(n) - 0.999999999999n + 0.499999999637 \log(n)} + \\ \frac{n}{O\left(\frac{e^{1.000000000000n \log(n) - 0.999999999999n + 0.499999999637 \log(n)}}{n^{2.000000000000}}\right)}$$

```
Mmx]
```


Number of alcohols: $s(z) = \frac{s(z)^3 + 2s(z^3)}{2}$

```
Mmx] use "symbolix"; use "multimix"; use "jorix"
```

```
Mmx] include "jorix/extrapolate_extra.mmx"
```

```
Mmx] bit_precision := 256; significant_digits := 15;
```

```
Mmx] s == fixed_point_series (f :->
      ((f*f*f + 2*dilate(f,3)) / 3) << 1, 1)
```

$$1 + z + z^2 + 2z^3 + 5z^4 + 11z^5 + 28z^6 + 74z^7 + 199z^8 + 551z^9 + O(z^{10})$$

```
Mmx] s[1000]
```

```
71414058382608563027996269715735039586933166020560927842446844\
19837729935282611084622722156252268983783693760331031649380178\
36695618256644646226261512335424552591745113792975199971801601\
25798467957908495421986885259041880858847239193631798561079741\
02892889269510412004093006951547101539486892379509940028587023\
90194705087308348927944684071303825503814190519839208648476765\
58240907916124487429778333667976863422529757274204058718557527\
28468762496469870238367567474487314760100550252695247565699736\
4188694861730348
```

```
Mmx] w == [ 1.0 * s[n] | n in 0..1000 ];
```

```
Mmx] guess_asymptotics (w, 9)
```

$$\begin{aligned} & 0.346304267394184 e^{1.19000938463478n - 1.50000000000000 \log(n)} - \\ & 0.146261408814197 e^{1.19000938463478n - 1.50000000000000 \log(n)} \\ & \frac{n}{n^{2.00000000000000}} + \\ & 0.138424784853633 e^{1.19000938463478n - 1.50000000000000 \log(n)} + \\ & \frac{n^{2.00000000000000}}{n^{3.00000000000000}} + \\ & 0.109383962532057 e^{1.19000938463478n - 1.50000000000000 \log(n)} + \\ & O\left(\frac{e^{1.19000938463478n - 1.50000000000000 \log(n)}}{n^{4.00000000000000}}\right) \end{aligned}$$

```
Mmx] exp (-1.19000938463478)
```

```
0.304218409074645
```

```
Mmx]
```



Transseries



```
Mmx] use "symbolix"; use "multimix"
```

```
Mmx] x == infinity ('x');
```

```
Mmx] lengthen (1 / (x - 1), 2)
```

$$\frac{1}{x} + \frac{1}{x^2} + \frac{1}{x^3} + \frac{1}{x^4} + \frac{1}{x^5} + \frac{1}{x^6} + O\left(\frac{1}{x^7}\right)$$

```
Mmx] 1 / (1 - 1/x - exp(-x))
```

$$1 + \frac{1}{x} + \frac{1}{x^2} + \frac{1}{x^3} + O\left(\frac{1}{x^4}\right) + \frac{1}{e^x} + \frac{2}{x e^x} + \frac{3}{x^2 e^x} + O\left(\frac{1}{x^3 e^x}\right) + \frac{1}{e^{2x}} + \frac{3}{x e^{2x}} + O\left(\frac{1}{x^2 e^{2x}}\right) + \frac{1}{e^{3x}} + O\left(\frac{1}{x e^{3x}}\right)$$

```
Mmx] 1 / ((1 - 1/x) * (1 - exp(-x))) - 1 / (1 - 1/x)
```

$$\frac{1}{e^x} + \frac{1}{x e^x} + \frac{1}{x^2 e^x} + O\left(\frac{1}{x^3 e^x}\right) + \frac{1}{e^{2x}} + \frac{1}{x e^{2x}} + O\left(\frac{1}{x^2 e^{2x}}\right) + \frac{1}{e^{3x}} + O\left(\frac{1}{x e^{3x}}\right)$$

```
Mmx] integrate (exp (x^2), x)
```

$$\frac{e^{x^2}}{2x} + \frac{e^{x^2}}{4x^3} + \frac{3e^{x^2}}{8x^5} + \frac{15e^{x^2}}{16x^7} + O\left(\frac{e^{x^2}}{x^9}\right)$$

```
Mmx] integrate (x^x, x)
```

$$\frac{e^{x \log(x)}}{\log(x)} - \frac{e^{x \log(x)}}{\log(x)^2} + \frac{e^{x \log(x)}}{\log(x)^3} - \frac{e^{x \log(x)}}{\log(x)^4} + O\left(\frac{e^{x \log(x)}}{\log(x)^5}\right) + \frac{e^{x \log(x)}}{x \log(x)^3} - \frac{3e^{x \log(x)}}{x \log(x)^4} + \frac{6e^{x \log(x)}}{x \log(x)^5} + O\left(\frac{e^{x \log(x)}}{x \log(x)^6}\right) + \frac{e^{x \log(x)}}{x^2 \log(x)^4} - \frac{e^{x \log(x)}}{x^2 \log(x)^5} + O\left(\frac{e^{x \log(x)}}{x^2 \log(x)^6}\right) + \frac{2e^{x \log(x)}}{x^3 \log(x)^5} + O\left(\frac{e^{x \log(x)}}{x^4 \log(x)^6}\right)$$

```
Mmx] lengthen (sum (log x, x), 7)
```

$$x \log(x) - x - \frac{\log(x)}{2} + \frac{1}{12x} - \frac{1}{360x^3} + \frac{1}{1260x^5} - \frac{1}{1680x^7} + \frac{1}{1188x^9} + O\left(\frac{1}{x^{11}}\right)$$

```
Mmx] euler == exp (-x) * integrate (exp (x) / x, x);
```

```
Mmx] lengthen (euler, 5)
```

$$\frac{1}{x} + \frac{1}{x^2} + \frac{2}{x^3} + \frac{6}{x^4} + \frac{24}{x^5} + \frac{120}{x^6} + \frac{720}{x^7} + \frac{5040}{x^8} + \frac{40320}{x^9} + O\left(\frac{1}{x^{10}}\right)$$

```
Mmx] eval (euler, x, 1000.0)
```

0.00100100200602412072

```
Mmx]
```



Given: $f_0, \dots, f_N$

Assume: $f_n = a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots$

Phase 0

$$f_n = a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots$$

Phase 1

$$\begin{aligned} f_n &= a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots \\ (\Delta f)_n = f_{n+1} - f_n &= \frac{-a_1}{n^2} + \frac{a_1 - 2a_2}{n^3} + \frac{3a_2 - a_1}{n^4} + \dots \end{aligned}$$

Phase 2

$$\begin{aligned} f_n &= a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots \\ (\Delta f)_n &= \frac{-a_1}{n^2} + \frac{a_1 - 2a_2}{n^3} + \frac{3a_2 - a_1}{n^4} + \dots \\ (\Delta(n^2 \Delta f))_n &= \frac{2a_2 - a_1}{n^2} + \frac{3a_1 - 8a_2}{n^3} + \frac{23a_2 - 7a_1}{n^4} + \dots \end{aligned}$$

Phase 2-bis

$$\begin{aligned} f_n &= a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots \\ (\Delta f)_n &= \frac{-a_1}{n^2} + \frac{a_1 - 2a_2}{n^3} + \frac{3a_2 - a_1}{n^4} + \dots \\ (\Delta(n^2 \Delta f))_n &= \frac{2a_2 - a_1}{n^2} + \frac{3a_1 - 8a_2}{n^3} + \frac{23a_2 - 7a_1}{n^4} + \dots \\ n^2 \Delta(n^2 \Delta f) &\approx (n^2 \Delta(n^2 \Delta f))_{N-2} = c_2 \end{aligned}$$

Phase 1-bis

$$f_n = a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots$$

$$(\Delta f)_n = \frac{-a_1}{n^2} + \frac{a_1 - 2a_2}{n^3} + \frac{3a_2 - a_1}{n^4} + \dots$$

$$(\Delta(n^2 \Delta f))_n = \frac{2a_2 - a_1}{n^2} + \frac{3a_1 - 8a_2}{n^3} + \frac{23a_2 - 7a_1}{n^4} + \dots$$

$$n^2 \Delta(n^2 \Delta f) \approx c_2$$

$$n^2 \Delta f \approx [n^2 \Delta f]_{N-1} + \Delta_{N-1}^{-1} \left(\frac{c_2}{n^2} \right) = c_1 - \frac{c_2}{n} - \frac{c_2}{2n^2} - \frac{c_2}{6n^3} + \dots$$

Phase 0-bis

$$f_n = a_0 + \frac{a_1}{n} + \frac{a_2}{n^2} + \dots$$

$$(\Delta f)_n = \frac{-a_1}{n^2} + \frac{a_1 - 2a_2}{n^3} + \frac{3a_2 - a_1}{n^4} + \dots$$

$$(\Delta(n^2 \Delta f))_n = \frac{2a_2 - a_1}{n^2} + \frac{3a_1 - 8a_2}{n^3} + \frac{23a_2 - 7a_1}{n^4} + \dots$$

$$n^2 \Delta(n^2 \Delta f) \approx c_2$$

$$n^2 \Delta f \approx c_1 - \frac{c_2}{n} - \frac{c_2}{2n^2} - \frac{c_2}{6n^3} + \dots$$

$$f \approx f_N + \Delta_N^{-1} \left(\frac{c_1}{n^2} - \frac{c_2}{n^3} - \dots \right) = c_0 - \frac{c_1}{n} + \frac{c_2 - c_1}{2n^2} + \frac{4c_2 - c_1}{6n^3} + \dots$$

See also: E-algorithm



Constant tactic



Assumption:

$$f_n = c + R_n, \quad R_n \prec 1$$

Criterion:

$$\varepsilon_{f, f_N} < \delta$$

$$\varepsilon_{f, \varphi} = \max_{k \in \{L, \dots, N\}} \frac{|f_n - \varphi_n|}{|f_n| + |\varphi_n|}$$

Transformation:

$$f \longmapsto g = \Delta f$$

Inverse transformation:

$$\begin{aligned} g_n &\approx \tilde{g}_n \\ f_n &\approx \tilde{f}_n = \left[f_N - \sum_{0 \leq i < N} \tilde{g}_i \right] + \sum_{0 \leq i < n} \tilde{g}_i \end{aligned}$$



Explicit tactic



Assumption:

$$f_n = \psi_n(c + R_n), \quad R_n \prec 1$$

Criterion:

$$\varepsilon_{f, \psi} f_N / \psi_N < \delta_\psi$$

$$\varepsilon_{f, \varphi} = \max_{k \in \{L, \dots, N\}} \frac{|f_n - \varphi_n|}{|f_n| + |\varphi_n|}$$

Transformation:

$$f \longmapsto g = \Delta(f/\psi)$$

Inverse transformation:

$$g_n \approx \tilde{g}_n$$

$$f_n \approx \tilde{f}_n = \psi_n \left(\left[f_N / \psi_N - \sum_{0 \leq i < N} \tilde{g}_i \right] + \sum_{0 \leq i < n} \tilde{g}_i \right)$$



Exponential tactic



Assumption:

$$f_n = \pm e^{R_n} \quad (R_n \succ 1).$$

Criterion:

$$\frac{f_n}{f_N} > 0 \quad (n = L, \dots, N),$$

Transformation:

$$f \longmapsto g = \log \frac{f}{\text{sign } f_N}$$

Inverse transformation:

$$\begin{aligned} g_n &\approx \tilde{g}_n \\ f_n &\approx \tilde{f}_n = (\text{sign } f_N) \exp \tilde{g}_n \end{aligned}$$



Asymptotic extrapolation



Combining tactics

Several tactics with transformations $\Phi_1, \dots, \Phi_p$ and inverses $\tilde{\Phi}_1, \dots, \tilde{\Phi}_p$

Compute all valid $\tilde{f}_{i_1, \dots, i_l} = (\tilde{\Phi}_{i_1} \circ \dots \circ \tilde{\Phi}_{i_l} \circ \Phi_{i_l} \circ \dots \circ \Phi_{i_1})(f)$ until specified l

Return best $\tilde{f}_{i_1, \dots, i_l}$

Inverse transformations using transseries

Inverse transformations Δ^{-1} and $\exp$ done on “transseries”

Numerical evaluations done using “summing to the least term”