Fraction-free Computation of Simultaneous Padé Approximants

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Simultaneous Padé Approximants

Given power series $A_1(z), ..., A_m(z)$ and integers $n_0, ..., n_m$ Find polynomials $U_1(z), ..., U_m(z)$ and V(z) with :

$$\qquad \qquad \blacktriangle_1(z) \approx \tfrac{U_1(z)}{V(z)}, \quad \dots, \quad A_m(z) \approx \tfrac{U_m(z)}{V(z)}.$$

▶ Degree bounds for $U_1(z), ..., U_m(z), V(z)$

Simultaneous Padé Approximants (more precise)

Given power series $A_1(z), ..., A_m(z)$ and integers $n_0, ..., n_m$ Find polynomials $U_1(z), ..., U_m(z)$ and V(z) with:

- ▶ $A_i(z) \approx \frac{U_i(z)}{V(z)}$ means $A_i(z)V(z) U_i(z) = \text{mod } z^{N+1}$
- ▶ Degree bounds for $U_1(z), ..., U_m(z), V(z)$

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$$\deg\left(U_{i}\right)\leq N-n_{i}\quad \deg\left(V\right)\leq N-n_{0},$$

where $N = n_0 + \cdots + n_m$.

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▶ Degree bounds for $U_1(z), ..., U_m(z), V(z)$

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where
$$N = n_0 + \cdots + n_m$$
.

Important fact: Closely related to Hermite-Padé approximants.



Example : e^z and e^{2z} and e^{3z}

$$e^{z} = \frac{-2z^{2} + 6z - 6}{6z^{3} - 11z^{2} + 12z - 6} + O(z^{4})$$

$$e^{2z} = \frac{z^{2} - 6}{6z^{3} - 11z^{2} + 12z - 6} + O(z^{4})$$

$$e^{3z} = \frac{-2z^{2} - 6z - 6}{6z^{3} - 11z^{2} + 12z - 6} + O(z^{4})$$

Here
$$(n_0, n_1, n_2, n_3) = (0, 1, 1, 1)$$
 so $N = 3$.

Degree bounds : denom = 3, numers = 2, 2, 2, order = 4.



Where used?

- ▶ Transcendence of important numbers : e.g. e, π , etc
- Inversion formulae for structured matrices
- Vector rational reconstruction problem:

$$\mathbb{K}^n(z) \to \mathbb{K}^n[[z]]$$
 - solve problem - $\mathbb{K}^n[[z]] \to \mathbb{K}^n(z)$
(e.g. linear system solving)

etc.

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People

- ► 1850-1890: Hermite
 - used these (informally) to prove transcendence of e (1873)
- ▶ 1890s: H. Padé
 - student of Hermite
 - first systematic study of rational approximation (1893)
 - Padé table
- 1920-1967: K. Mahler
 - a formal study in general case
 - relationships between groups of simultaneous Padé approximants and Hermite-Padé approximants

How to Compute Simultaneous Padé Approximants

- linear system of equations for coefficients
- represent as a vector Hermite-Padé problem

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Last two methods used to find *all* solutions to approximation problem. Here $all \equiv basis$ as a module

All three methods work with Fraction-Free arithmetic.

Linear System : A(z) and B(z) and C(z)

Can always solve via linear system for coefficients

$$(a_0+a_1z+\cdots)(v_0+v_1z+v_2z^2+v_3z^3)-(u_0+u_1z+u_2z^2)=O(z^4)$$

$$(b_0+b_1z+\cdots)(v_0+v_1z+v_2z^2+v_3z^3)-(\hat{u}_0+\hat{u}_1z+\hat{u}_2z^2)=O(z^4)$$

$$(c_0+c_1z+\cdots)(v_0+v_1z+v_2z^2+v_3z^3)-(u_0^*+u_1^*z+u_2^*z^2)=O(z^4)$$

Notice that matrix is highly structured.



Simultaneous-Padé as Vector Hermite-Padé

$$A(z)U_0(z) - U_1(z) = O(z^4)$$

 $B(z)U_0(z) - U_2(z) = O(z^4)$
 $C(z)U_0(z) - U_3(z) = O(z^4)$

same as

$$\left[\begin{array}{c}A(z)\\B(z)\\C(z)\end{array}\right]U_0(z)+\left[\begin{array}{c}-1\\0\\0\end{array}\right]U_1(z)+\left[\begin{array}{c}0\\-1\\0\end{array}\right]U_2(z)+\left[\begin{array}{c}0\\0\\-1\end{array}\right]U_3(z)=0(z^4)$$

Computation

- Hermite-Padé (general):
 - Beckermann-Labahn (SIMAX 1994)
 - P. Giorgi, C-P. Jeannerod, G. Villard (ISSAC 2003)
 - fast polynomial matrix arithmetic
 - Beckermann-Labahn (SIMAX 2000)
 - fraction-free arithmetic
 - Zhou-Labahn (Next talk)
 - fast arithmetic
- Simultaneous-Padé
 - Olesky and Storjohann (2007)

Cost

If input has size $O(\kappa)$ then :

- Fraction-Free Gaussian Elimination (FFGE) :
 - Bit complexity of operations: $O(\kappa^2 m^6 N^5)$
- B-L [SIMAX 2000] :
 - Bit complexity of operations: $O(\kappa^2 m^5 N^4)$
 - Size of objects : $O(\kappa mN)$
- Today
 - Bit complexity of operations: $O(\kappa^2 m^2 N^4)$
 - Size of objects : $O(\kappa mN)$

Order Basis

Input : Vector of power series $\mathbf{A}(z) \in \mathbb{K}^{1 \times m}[z]$ and integer σ Order Basis : Matrix polynomial $\mathbf{M}(z) \in \mathbb{K}m \times m[z]$

$$\mathbf{A}(z)\cdot\mathbf{M}(z)=z^{\sigma}\mathbf{R}(z)$$

▶ All solutions $\mathbf{V}(z) \in \mathbb{K}^{m \times 1}[z]$ of $\mathbf{A}(z) \cdot \mathbf{V}(z) = O(z^{\sigma})$ is a combination of columns of $\mathbf{M}(z)$.

$$\mathbf{V}(z) = \alpha_1(z)\mathbf{M}^{(1)}(z) + \cdots + \alpha_m(z)\mathbf{M}^{(m)}(z)$$

▶ degree constraints become degree constraints on $\alpha_i(z)$.



BL [SIMAX 2000]

For given vector of power series and given vector of degree bounds \vec{n} it computes an order basis for problem

$$A_0(z)U_0(z) + \cdots + A_m(z)U_m(z) = O(z^{N+1})$$

with degree bounds

$$\deg(U_i) \leq n_i - 1.$$

- Order bases for Hermite-Padé approximant problem
- Fraction-free method (FFFG)
- Works for vector and matrix power series

Simultaneous-Padé

Solve as

$$\begin{bmatrix} A_1(z) \\ A_2(z) \\ \vdots \\ A_m(z) \end{bmatrix} U_0(z) + \begin{bmatrix} -A_0(z) \\ 0 \\ \vdots \\ 0 \end{bmatrix} U_1(z) + \begin{bmatrix} 0 \\ -A_0(z) \\ \vdots \\ 0 \end{bmatrix} U_2(z) + \cdots + \begin{bmatrix} 0 \\ 0 \\ \vdots \\ -A_0(z) \end{bmatrix} U_m(z) = O(z^{N+1}).$$

Simultaneous-Padé becomes vector Hermite-Padé problem.

Use FFFG to solve in exact arithmetic.

Duality

Right and Left Matrix Padé duality:

$$A(z)V(z) - U(z) = O(z^K)$$
 and $\hat{V}(z)A(z) - \hat{U}(z) = O(z^K)$

Hermite-Padé and Simultaneous-Padé duality:

$$A_0(z)U_0(z) + [A_1(z), \cdots, A_m(z)] \begin{bmatrix} U_1(z) \\ \vdots \\ U_m(z) \end{bmatrix} = O(z^{N+1})$$

$$V(z)[A_1(z), \cdots, A_m(z)] - [U_1(z), \cdots, U_m(z)] A_0(z) = O(z^{(N+1, \cdots, N+1)})$$

Useful for inversion formulae (L [LAA 1992])



Duality

Right and Left Matrix Padé duality:

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Hermite-Padé and Simultaneous-Padé duality:

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Useful for inversion formulae (L [LAA 1992])



Duality: Much Better

Hermite-Padé and Simultaneous-Padé order bases are duals to each other (B-L [JCAM 1997]) .

$$\mathbf{A}(z)\mathbf{M}(z) = \begin{bmatrix} A_0(z) & A_1(z) & \cdots & A_m(z) \\ 0 & 1 & & & \\ \vdots & & & \ddots & \\ 0 & & & 1 \end{bmatrix} \begin{bmatrix} U_{00}(z) & \cdots & U_{0,m}(z) \\ \vdots & & & \vdots \\ \vdots & & & \vdots \\ U_{m0}(z) & \cdots & U_{m,m}(z) \end{bmatrix} = O(z^{(N+1,0,\cdots,0)})$$

$$\mathbf{A}^*(z)\mathbf{M}^*(z) \left[\begin{array}{cccc} 1 & 0 & \cdots & 0 \\ -A_1(z) & A_0(z) & & & \\ \vdots & & \ddots & & \\ \vdots & & & \ddots & \\ -A_m(z) & & & A_0(z) \end{array} \right] \left[\begin{array}{cccc} V_{00}(z) & \cdots & V_{0,m}(z) \\ \vdots & & & \vdots \\ \vdots & & & \vdots \\ V_{m0}(z) & \cdots & V_{m,m}(z) \end{array} \right] = O(z^{(0,N+1,\cdots,N+1)})$$

where $\mathbf{P}^*(z) = \operatorname{adj}(\mathbf{P}(z))^T = \operatorname{cof}(\mathbf{P}(z))$.

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Computation Process

Input : $\mathbf{A}(z)$, \vec{n} .

Process for Hermite-Padé : Computes Order bases of type

$$\vec{v}^{(0)}, \vec{v}^{(1)}, \ldots, \vec{v}^{(N+1)}$$

$$\mathbf{I} = \mathbf{M}(\vec{v}^{(0)}, z) \quad \rightarrow \quad \mathbf{M}(\vec{v}^{(1)}, z) \rightarrow \quad \cdots \quad \rightarrow \mathbf{M}(\vec{v}^{(N+1)}, z)$$

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$$\mathbf{I} = \mathbf{M}^*(\vec{v}^{(0)}, z) \quad \rightarrow \quad \mathbf{M}^*(\vec{v}^{(1)}, z) \rightarrow \quad \cdots \quad \rightarrow \mathbf{M}^*(\vec{v}^{(N+1)}, z)$$



Computation Process

Input : $\mathbf{A}(z)$, \vec{n} .

Process for Hermite-Padé : Computes Order bases of type

$$\vec{v}^{(0)}, \vec{v}^{(1)}, \ldots, \vec{v}^{(N+1)}$$

$$\begin{split} \mathbf{I} &= \mathbf{M}(\vec{v}^{(0)},z) \quad \rightarrow \quad \mathbf{M}(\vec{v}^{(1)},z) \rightarrow \quad \cdots \quad \rightarrow \mathbf{M}(\vec{v}^{(N+1)},z) \\ \uparrow \qquad \qquad \uparrow \qquad \qquad \uparrow \qquad \qquad \downarrow \\ \downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow \\ \mathbf{I} &= \mathbf{M}^*(\vec{v}^{(0)},z) \quad \rightarrow \quad \mathbf{M}^*(\vec{v}^{(1)},z) \rightarrow \quad \cdots \quad \rightarrow \mathbf{M}^*(\vec{v}^{(N+1)},z) \end{split}$$

Recursion (Hermite-Padé)

$$\mathbf{M}(\vec{v}^{(i+1)},z) = \mathbf{M}(\vec{v}^{(i)},z)\mathbf{A}(z)\mathbf{B}(z)$$

$$\begin{bmatrix} m_{11}^{(i+1)}(z) & \cdots & m_{1,m}^{(i+1)}(z) \\ \vdots & & \vdots \\ m_{m1}^{(i+1)}(z) & \cdots & m_{m,m}^{(i+1)}(z) \end{bmatrix} = \begin{bmatrix} m_{11}^{(i)}(z) & \cdots & m_{1,m}^{(i)}(z) \\ \vdots & & \vdots \\ m_{m1}^{(i)}(z) & \cdots & m_{m,m}^{(i)}(z) \end{bmatrix} \begin{bmatrix} * & * & * \\ * & * & * \\ & \ddots & & * \end{bmatrix} \begin{bmatrix} z & & & & \\ * & * & & \\ * & * & * \end{bmatrix}$$

Recursion (Simultaneous-Padé)

Take adjoints and then transposes to get

$$\mathbf{M}^*(\vec{v}^{(i+1)}, z) = \mathbf{M}^*(\vec{v}^{(i)}, z)\mathbf{A}^*(z)\mathbf{B}^*(z)$$

$$\left[\begin{array}{ccc} \hat{m}_{11}^{(i+1)}(z) & \cdots & \hat{m}_{1,m}^{(i+1)}(z) \\ \vdots & & \vdots \\ \hat{m}_{m1}^{(i+1)}(z) & \cdots & \hat{m}_{m,m}^{(i)}(z) \end{array} \right] = \left[\begin{array}{ccc} \hat{m}_{11}^{(i)}(z) & \cdots & \hat{m}_{1,m}^{(i)}(z) \\ \vdots & & \vdots \\ \hat{m}_{m1}^{(i)}(z) & \cdots & \hat{m}_{m,m}^{(i)}(z) \end{array} \right] \left[\begin{array}{ccc} * & * & * \\ 0 & * & * \\ \vdots & & \vdots \\ 0 & * & * \end{array} \right] \left[\begin{array}{cccc} 1 & & & & \\ z & & & \\ \vdots & & & \vdots \\ 0 & * & * \end{array} \right]$$

The Algorithm

At each iteration for Hermite-Padé

- lacktriangleright Increase order of each row using 'special' pivot column π
- ▶ Increase order of 'special' pivot column π
- Normalize order basis to get special shifted Popov form

At each iteration for Simultaneous-Padé

- Increase order of 'special' row using fraction-free Gaussian elimination on first term of residual
- Increase order of 'special' pivot row
- Normalize order basis to get special shifted Popov form

Future Research

- ► Fraction-Free → modular methods
- Duality for alternative order basis algorithm for Hermite-Padé and vector Hermite-Padé approximation problem
- Use alternative order basis algorithm for noncommutative case of Ore matrix polynomials
- Use above algorithm to create faster algorithms for matrix polynomial and matrix Ore normal forms (Popov, Hermite)