

Identifying the non-trivial zeros of the Riemann zeta function for prime counting function approximation in the Loewner framework

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C. Poussot-Vassal, I.V. Gosea, P. Vuillemin and A.C. Antoulas

[Onera / Max Planck Institute / Rice University]
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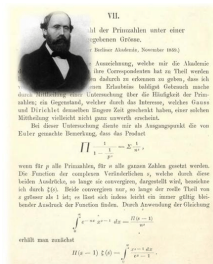
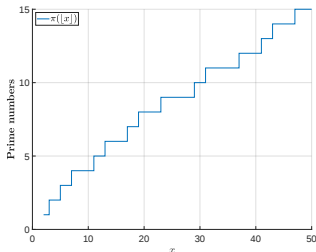


Introduction

Glimpse of the Riemann conjecture idea

B. Riemann made an important (and unexpected) connection between number theory and complex functions analysis

$$\pi(\lfloor x \rfloor)$$

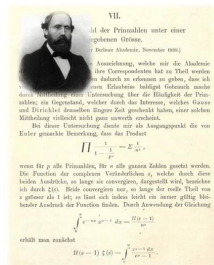
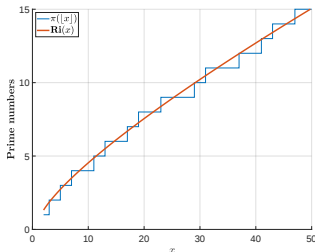


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B. Riemann made an important (and unexpected) connection between number theory and complex functions analysis

$$\pi(\lfloor x \rfloor) \approx \text{Ri}(x)$$



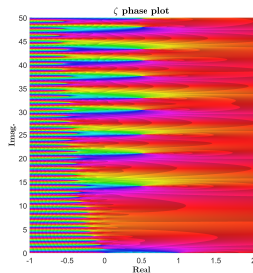
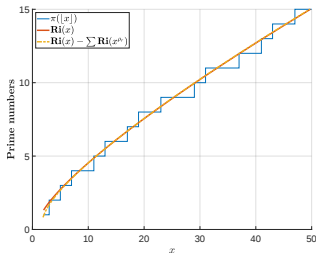
$$\text{Ri}(x) = \sum_{n=1}^{\infty} \frac{\mu(n)}{n} \text{Li}(x^{1/n}), \mu(n) \text{ is the Mobius function and } \text{Li}(x) = \int_2^x \frac{dt}{\ln(t)}$$

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$$\pi(\lfloor x \rfloor) \approx \text{Ri}(x) - \sum_{\rho_r} \text{Ri}(x^{\rho_r}) \quad (\text{trivial zeros})$$



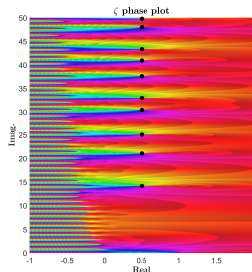
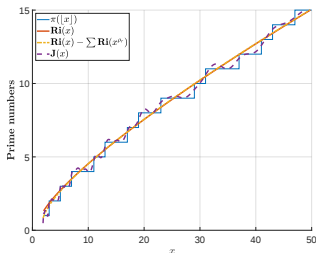
$$\text{Ri}(x) = \sum_{n=1}^{\infty} \frac{\mu(n)}{n} \text{Li}(x^{1/n}), \quad \mu(n) \text{ is the Mobius function and } \text{Li}(x) = \int_2^x \frac{dt}{\ln(t)}$$

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B. Riemann made an important (and unexpected) connection between number theory and complex functions analysis

$$\pi(\lfloor x \rfloor) \approx \mathbf{Ri}(x) - \sum_{\rho_r} \mathbf{Ri}(x^{\rho_r}) - \sum_{\rho_i} \mathbf{Ri}(x^{\rho_i}) = \mathbf{J}(x) \text{ (trivial and non-trivial zeros)}$$



$$\mathbf{Ri}(x) = \sum_{n=1}^{\infty} \frac{\mu(n)}{n} \mathbf{Li}(x^{1/n}), \mu(n) \text{ is the Mobius function and } \mathbf{Li}(x) = \int_2^x \frac{dt}{\ln(t)}$$

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Glimpse of the Riemann conjecture idea¹

$$\mathbf{J}(x) = \mathbf{Ri}(x) - \sum_{\rho_r} \mathbf{Ri}(x^{\rho_r}) - \sum_{\rho_i} \mathbf{Ri}(x^{\rho_i})$$

¹Authors warmly thank [David Louapre](https://scienceetonnante.com/) (<https://scienceetonnante.com/>) for providing the Python code allowing to compute the Riemann prime counting function.

Introduction

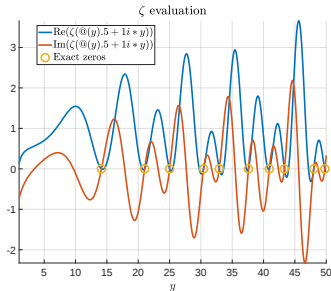
The ζ function... a central element

Initially defined as ($z \in \mathbb{C}_{\Re(z)>1}$)

$$\zeta(z) = \sum_{n=1}^{\infty} \frac{1}{n^z}$$

A continuation is used ($z \in \mathbb{C}_{\Re(z) \neq 1}$)

$$\zeta(z) = \frac{1}{1-2^{1-z}} \sum_{n=1}^{\infty} \frac{(-1)^{n-1}}{n^z}$$



Evaluation of ζ along

$$z = \frac{1}{2} + iy$$



Introduction

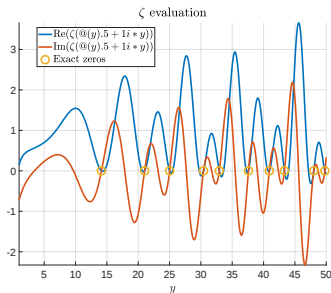
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Barycentric form

$$\hat{\zeta}(z) = \frac{\sum_i \beta_i / (z - \lambda_i)}{\sum_i \alpha_i / (z - \lambda_i)} \approx \zeta(z)$$



Introduction

This talk

Loewner and AAA to approximate ζ zeros

- ▶ Interpolatory **Loewner** method [Mayo/Antoulas]
- ▶ Interpolatory **AAA** method [Anderson/Antoulas] & [Trefethen]

We consider transfer function

$$\mathbf{H}(z) = \frac{\sum_i \beta_i / (z - \lambda_i)}{\sum_i \alpha_i / (z - \lambda_i)}$$

Realizations

$$\mathcal{S} : (E, A, B, C, \mathbf{0})$$

and Loewner pencil

$$(\mathbb{L}, \mathbb{M})$$

 A.C. Antoulas, C. Beattie and S. Gugercin, "*Interpolatory methods for model reduction*", SIAM Computational Science and Engineering, Philadelphia, 2020.

 A.J. Mayo and A.C. Antoulas, "*A framework for the solution of the generalized realization problem*", Linear Algebra and its Applications, 2007.

 Y. Nakatsukasa, O. Sete and L.N. Trefethen, "*The AAA algorithm for rational approximation*", SIAM Journal on Scientific Computing, 2018.

 L.N. Trefethen, "*Rational functions (von Neumann Prize lecture)*", SIAM Annual Meeting, 2020.

Interpolatory framework

Loewner model-based approximation (rational interpolation)

Given model $\mathbf{H}$ and μ_j ,
 λ_i , seek $\hat{\mathbf{H}}$ s.t.

$$\begin{aligned}\hat{\mathbf{H}}(\mu_j) &= \mathbf{H}(\mu_j) \\ \hat{\mathbf{H}}(\lambda_i) &= \mathbf{H}(\lambda_i)\end{aligned}$$

$$i = 1, \dots, k; j = 1, \dots, q.$$



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$$\mathbb{L} = \begin{bmatrix} \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1)}{\mu_1 - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k)}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1)}{\mu_q - \lambda_1} & \dots & \frac{\mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k)}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbb{M} = \begin{bmatrix} \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_1 - \lambda_1} & \dots & \frac{\mu_1 \mathbf{H}(\mu_1) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_1 - \lambda_k} \\ \vdots & \ddots & \vdots \\ \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_1) \lambda_1}{\mu_q - \lambda_1} & \dots & \frac{\mu_q \mathbf{H}(\mu_q) - \mathbf{H}(\lambda_k) \lambda_k}{\mu_q - \lambda_k} \end{bmatrix}$$

$$\mathbf{W} = \begin{bmatrix} \mathbf{H}(\lambda_1) & \dots & \mathbf{H}(\lambda_k) \end{bmatrix}$$

$$\mathbf{V}^T = \begin{bmatrix} \mathbf{H}(\mu_1) & \dots & \mathbf{H}(\mu_q) \end{bmatrix}$$

$$\hat{\mathbf{H}}(z) = \mathbf{W}(-z\mathbb{L} + \mathbb{M})^{-1}\mathbf{V} \Rightarrow \text{Rational interpolation}$$



Interpolatory framework

Loewner data-driven approximation (rational interpolation)

Given $\{\mu_j, \mathbf{v}_j\}$, $\{\lambda_i, \mathbf{w}_i\}$
data, seek $\hat{\mathbf{H}}$, s.t.

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$$\hat{\mathbf{H}}(z) = \mathbf{W}(-z\mathbb{L} + \mathbb{M})^{-1}\mathbf{V} \Rightarrow \text{Rational interpolation (data-driven)}$$



Interpolatory framework

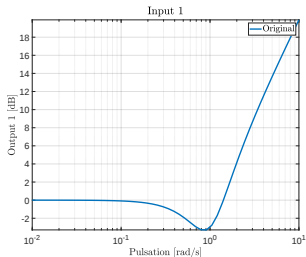
Some features of the Loewner landmark

Rational function satisfies

$$\hat{\mathbf{H}}(\lambda_i) = \mathbf{H}(\lambda_i) \text{ and } \hat{\mathbf{H}}(\mu_j) = \mathbf{H}(\mu_j)$$

$$\mathbf{H}(s) = \frac{s^2 + s + 1}{s + 1}$$

Sampled with $\lambda_i = [1, 2, \dots, 20]$ and $\mu_j = [1.5, 2.5, \dots, 20.5]$, leads to realization of dimension



Interpolatory framework

Some features of the Loewner landmark

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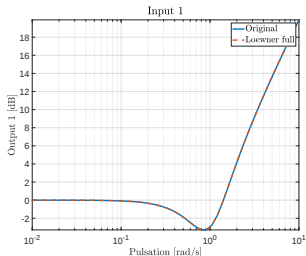
$$\hat{\mathbf{H}}(\lambda_i) = \mathbf{H}(\lambda_i) \text{ and } \hat{\mathbf{H}}(\mu_j) = \mathbf{H}(\mu_j)$$

Realization

$$\hat{\mathbf{H}}(s) = \mathbf{W}(-s\mathbf{L} + \mathbf{M})^{-1}\mathbf{V}$$

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Interpolatory framework

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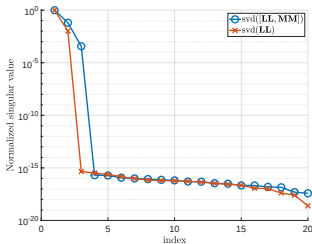
Rank reveals the underlying rational (r) and McMillan (ν) orders

$$\text{rank}([\mathbb{L}, \mathbb{M}]) = r$$

$$\text{rank}(\mathbb{L}) = \nu$$

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Interpolatory framework

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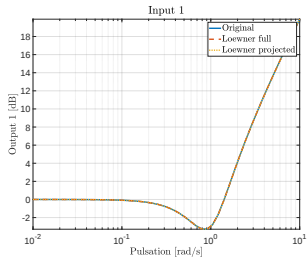
$$\text{rank}(\mathbb{L}) = \nu$$

Minimal realization

$$\hat{\mathbf{H}}_r(s) = \mathbf{W}X(-s\mathbf{Y}^T\mathbb{L}X + \mathbf{Y}^T\mathbb{M}X)^{-1}\mathbf{Y}^T\mathbf{V}$$

$$\mathbf{H}(s) = \frac{s^2 + s + 1}{s + 1}$$

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Interpolatory framework

AAA model-based approximation (rational interpolation)

Given $\{\nu_j, h_j\}$, $\{\eta_i, g_i\}$
data, seek $\hat{\mathbf{H}}$, s.t.

$$\hat{\mathbf{H}}_\ell(\xi) = \frac{\sum_{j=1}^{\ell} \frac{\alpha_j^{(\ell)} h_j}{\xi - \nu_j}}{\sum_{j=1}^{\ell} \frac{\alpha_j^{(\ell)}}{\xi - \nu_j}}$$

$i = 1, \dots, \ell$ and $j = 1, \dots, 2n - \ell$.

Mixes **Least Squares** with **Loewner** in an **iteration**

- ▶ Find $\nu_\ell \in \Omega^{(\ell-1)}$, where
 $\nu_\ell = \arg \max_{1 \leq k \leq 2M} \|f_k - \hat{\mathbf{H}}_{\ell-1}(z_k)\|$
- ▶ Set $\Omega^{(\ell)} := \Omega^{(\ell-1)} \setminus \{\nu_\ell\}$, where ,
 $h^\ell := \mathbf{H}(\nu_\ell)$
- ▶ Compute $\alpha^{(\ell)} = -\mathbb{L}^\# \mathbf{f}$
- ▶ Set

$$\begin{aligned}\hat{E} &= \text{diag}(1, \dots, 1, 0) \\ \hat{A} &= \text{diag}(\nu_1, \dots, \nu_r, 1) - \hat{B} \mathbf{e}_{r+1}^T \\ \hat{B} &= [\alpha_1^{(r)}, \dots, \alpha_r^{(r)}, 1]^T \\ \hat{C} &= [h_1, \dots, h_r, 0]\end{aligned}$$



Y. Nakatsukasa, O. Sete and L.N. Trefethen, "*The AAA algorithm for rational approximation*", SIAM Journal on Scientific Computing, 2018.



Chebfun Developers, "<http://www.chebfun.org/>", University of Oxford.

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$\hat{\mathbf{H}}_\ell(z) = \hat{C}(z\hat{E} - \hat{A})^{-1}\hat{B} \Rightarrow$ best average fitting of the data (data-driven)



Y. Nakatsukasa, O. Sete and L.N. Trefethen, "*The AAA algorithm for rational approximation*", SIAM Journal on Scientific Computing, 2018.



Chebfun Developers, "<http://www.chebfun.org/>", University of Oxford.

Numerical illustration

Process

A continuation of ζ is used
($z \in \mathbb{C}_{\Re(z) \neq 1}$)

$$\frac{1}{1 - 2^{1-z}} \sum_{n=1}^{\infty} \frac{(-1)^{n-1}}{n^z}$$

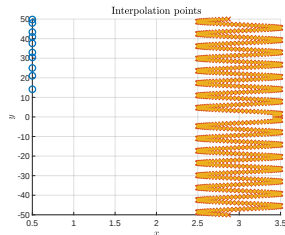
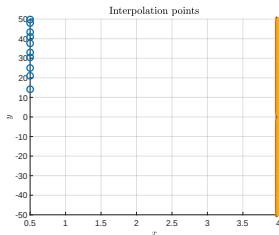


Rational approximation

ROM

- ▶ Rational function
- ▶ r state variables

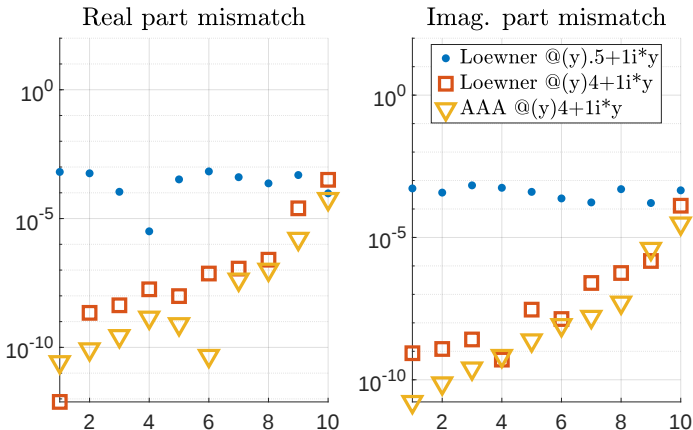
μ_j and λ_i (Loewner case) & ν_j and η_i (AAA case)
As chosen by Trefethen et al. (left) & Random case (right)



Numerical illustration

Zeros computation on the rational interpolant

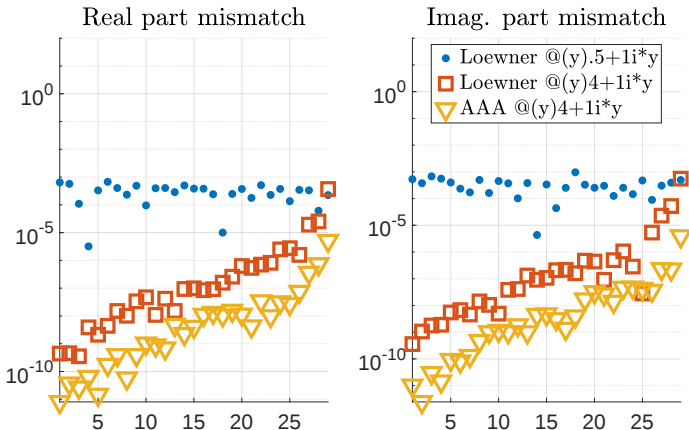
Interval $[0, 50]$, $n = 1000$, $\zeta(4 + iy)$



Numerical illustration

Zeros computation on the rational interpolant

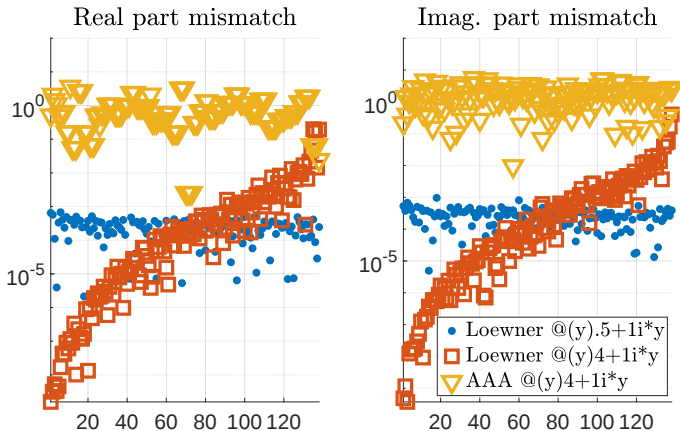
Interval $[0, 100]$, $n = 1000$, $\zeta(4 + iy)$



Numerical illustration

Zeros computation on the rational interpolant

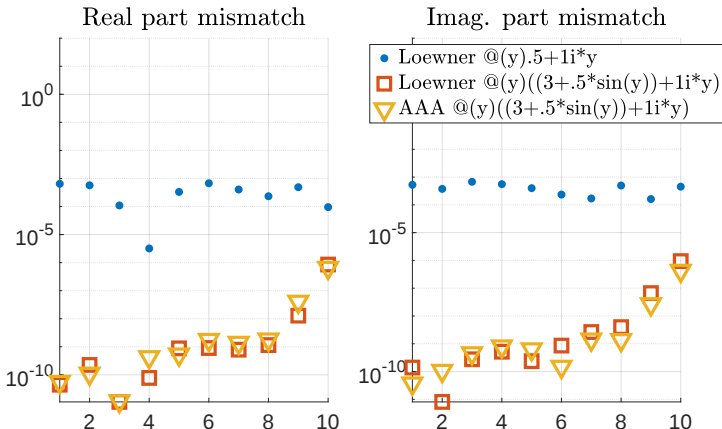
Interval $[0, 300]$, $n = 1000$, $\zeta(4 + iy)$



Numerical illustration

Zeros computation on the rational interpolant

Interval $[0, 50]$, $n = 1000$, $\zeta(3 + 0.5 \sin(y) + iy)$

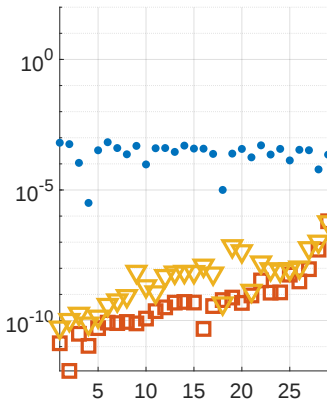


Numerical illustration

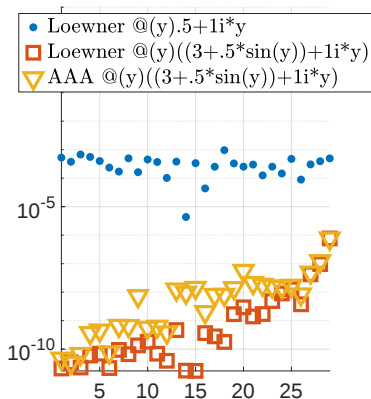
Zeros computation on the rational interpolant

Interval $[0, 100]$, $n = 1000$, $\zeta(3 + 0.5 \sin(y) + iy)$

Real part mismatch



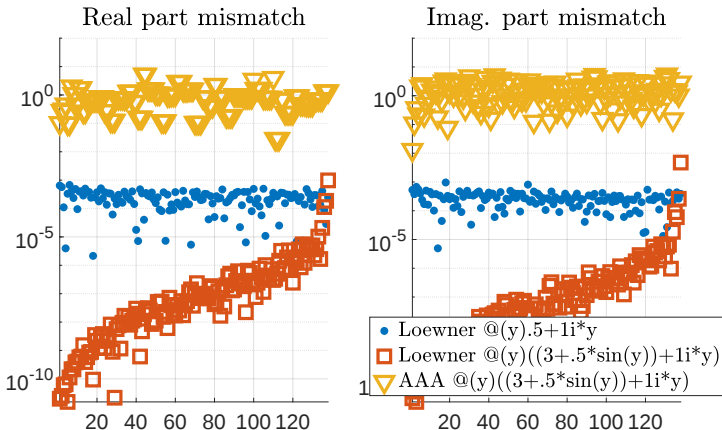
Imag. part mismatch



Numerical illustration

Zeros computation on the rational interpolant

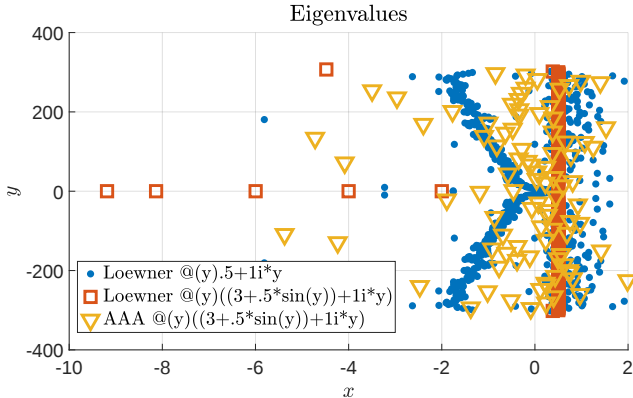
Interval $[0, 300]$, $n = 1000$, $\zeta(3 + 0.5 \sin(y) + iy)$



Numerical illustration

Zeros computation on the rational interpolant

Interval $[0, 300]$, $n = 1000$, $\zeta(3 + 0.5 \sin(y) + iy)$

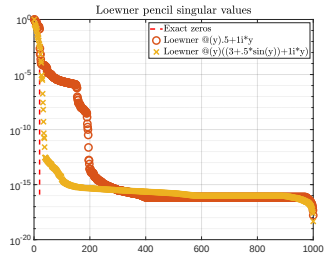
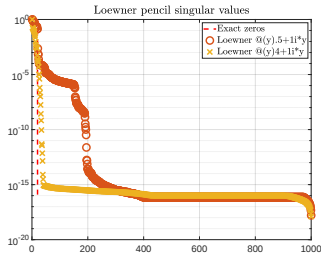


Loewner pencil finds trivial and non-trivial zeros, whatever the choice of interpolation points

Numerical illustration

Pencil singular values

Interval $[0, 50]$, $n = 1000$, $\zeta(4 + iy)$ & $\zeta(3 + 0.5 \sin(y) + iy)$

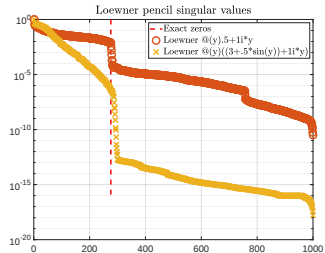
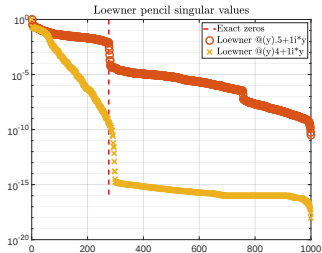


Loewner pencil detects the number of non-trivial zeros

Numerical illustration

Pencil singular values

Interval $[0, 300]$, $n = 1000$, $\zeta(4 + iy)$ & $\zeta(3 + 0.5\sin(y) + iy)$

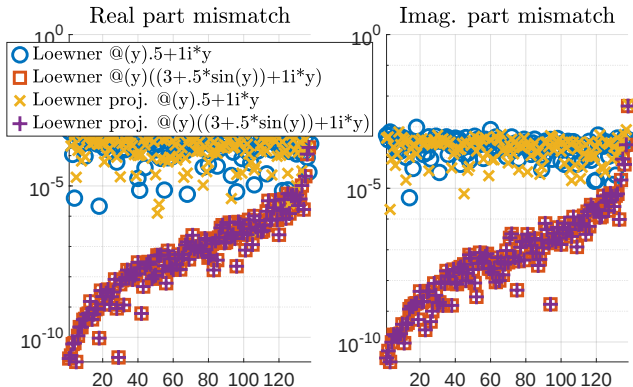


Loewner pencil detects the number of non-trivial zeros

Numerical illustration

Zeros computation with projected rational model

Interval $[0, 300]$, $n = 1000$, $\zeta(4 + iy)$ & $\zeta(3 + 0.5\sin(y) + iy)$



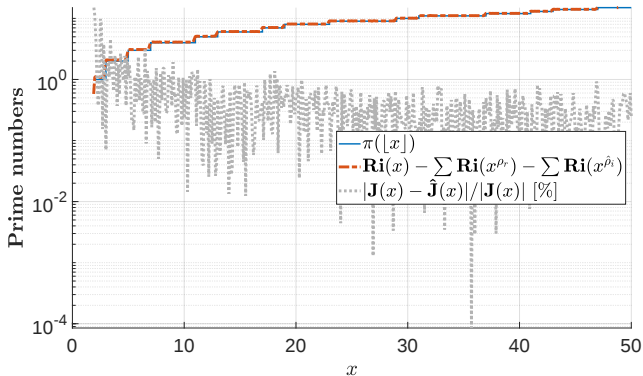
$$\hat{\mathbf{H}}_r(s) = \mathbf{W} \mathbf{X} (-s \mathbf{Y}^T \mathbf{L} \mathbf{X} + \mathbf{Y}^T \mathbf{M} \mathbf{X})^{-1} \mathbf{Y}^T \mathbf{V}$$

Minimal rational function catches the zeros as accurately as full realization!

Conclusions

Back to $\mathbf{J}(x) \dots \hat{\mathbf{J}}(x)$

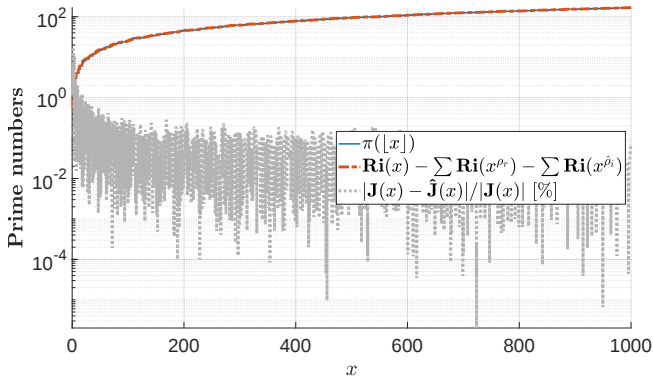
$$\hat{\mathbf{J}}(x) = \mathbf{Ri}(x) - \sum_{\rho_r} \mathbf{Ri}(x^{\rho_r}) - \sum_{\rho_i i=1 \dots 500} \mathbf{Ri}(x^{\hat{\rho}_i})$$



Conclusions

Back to $J(x)$... $\hat{J}(x)$

$$\hat{J}(x) = \text{Ri}(x) - \sum_{\rho_r} \text{Ri}(x^{\rho_r}) - \sum_{\rho_i i=1 \dots 200} \text{Ri}(x^{\hat{\rho}_i})$$



Conclusions

Take home message²

Loewner and **AAA** enforce rational approximation of complex functions,

- ▶ **Loewner** and **AAA** find "accurate" rational models
- ▶ **Loewner pencil** catches the exact number of ζ zeros at low cost
- ▶ **Loewner** seems more robust to the interpolation points selection

→ How to select interpolation points?

→ Any idea? Please contact us!

Bernhard Riemann (1826 - 1866)



Karl Loewner (1893 - 1968)



²Ack. to **David Louape** (<https://scienceetonnante.com/>) for inspiration and help.

Identifying the non-trivial zeros of the Riemann zeta function for prime counting function approximation in the Loewner framework

Curves and Surfaces 2022 (Arcachon, France)

C. Poussot-Vassal, I.V. Gosea, P. Vuillemin and A.C. Antoulas

[Onera / Max Planck Institute / Rice University]
June, 2022



Appendix

Interpolation points?

Interval $[0, 100]$, $n = 1000$, $\zeta(4 + iy)$ & $\zeta(3 + 0.5 \sin(y) + iy)$

Alternate vs. Random shift

