

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

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[Onera / Max Planck Institute / Rice University]  
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MAX PLANCK INSTITUTE  
FOR DYNAMICS OF COMPLEX  
TECHNICAL SYSTEMS  
MAGDEBURG

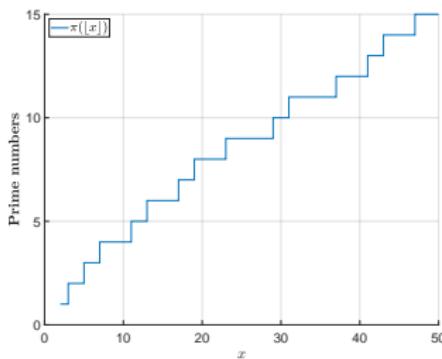


# Introduction

## Glimpse of the Riemann conjecture idea

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

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

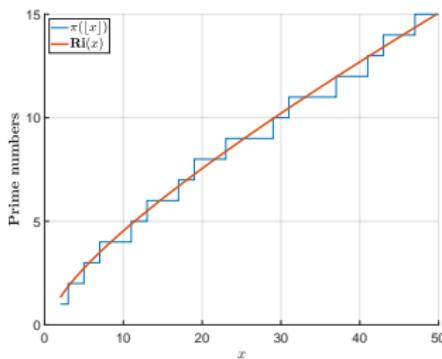


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$$\pi(\lfloor x \rfloor) \approx \mathbf{Ri}(x)$$



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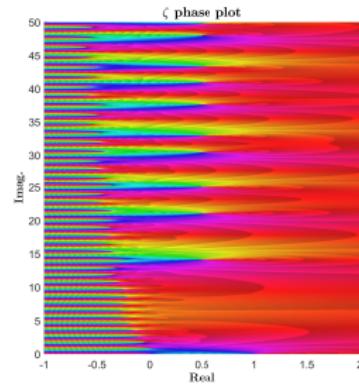
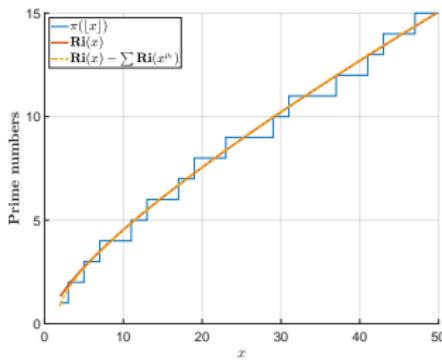
$$\mathbf{Ri}(x) = \sum_{n=1}^{\infty} \frac{\mu(n)}{n} \mathbf{Li}(x^{1/n}), \text{ where } \mu(n) \text{ is the Möbius function and } \mathbf{Li}(x) = \int_2^x \frac{dt}{\ln(t)}$$

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



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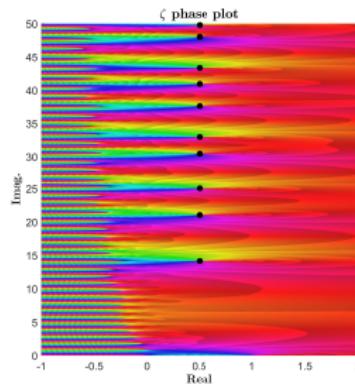
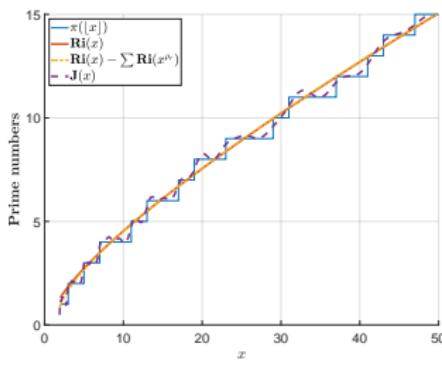
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$$\pi(\lfloor x \rfloor) \approx \text{Ri}(x) - \sum_{\rho_r} \text{Ri}(x^{\rho_r}) - \sum_{\rho_i} \text{Ri}(x^{\rho_i}) = \mathbf{J}(x) \text{ (trivial and non-trivial zeros)}$$



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

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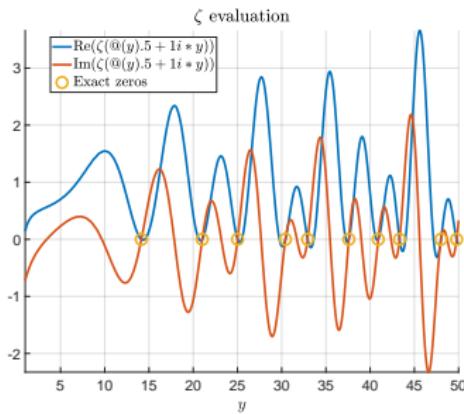
<sup>1</sup>Authors warmly thank **David Louapre** (<https://scienceetonnante.com/>) for providing the Python code allowing to compute the Riemann prime counting function.

# Introduction

The  $\zeta$  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  $\zeta$  along

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

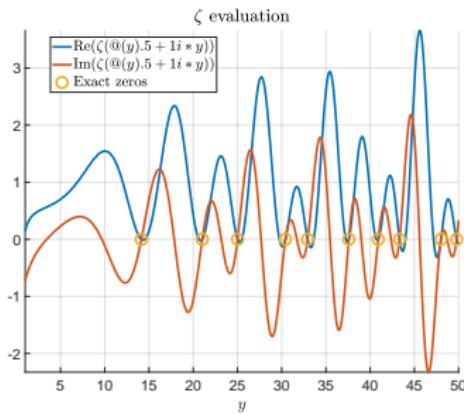


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Barycentric form

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



# Introduction

This talk

## Loewner and AAA to approximate $\zeta$ 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

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

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

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

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

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# Interpolatory framework

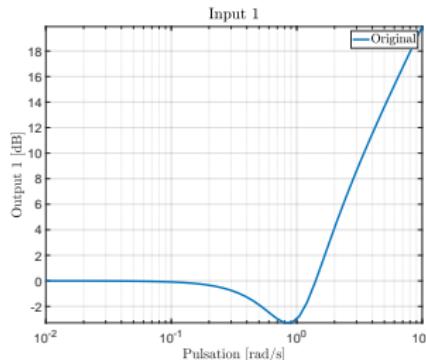
Some features of the Loewner landmark

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



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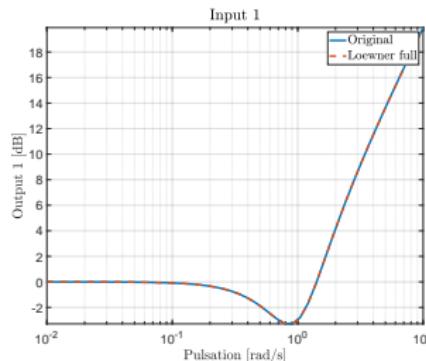
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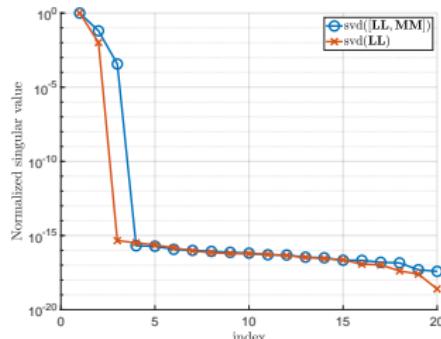
Rank reveals the underlying rational ( $r$ ) and McMillan ( $\nu$ ) orders

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

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

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Sampled with  $\lambda_i = [1, 2, \dots, 20]$  and  $\mu_j = [1.5, 2.5, \dots, 20.5]$ , leads to realization of dimension  $r = 3$  ( $\nu = 2$ )



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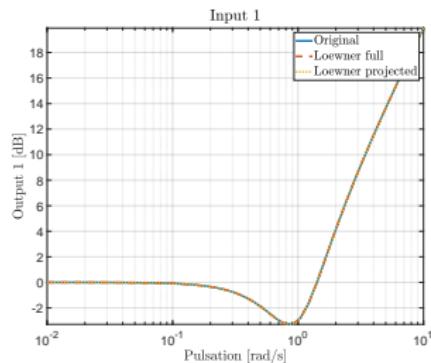
$$\text{rank}(\mathbb{L}) = \nu$$

Minimal realization

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

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

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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)

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# Numerical illustration

Process

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

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

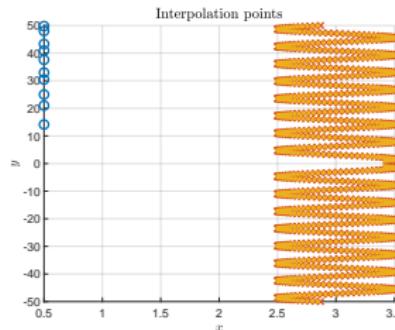
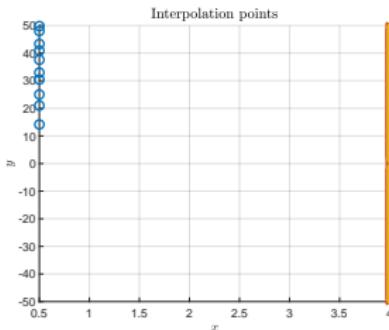


Rational approximation

ROM

- ▶ Rational function
- ▶  $r$  state variables

$\mu_j$  and  $\lambda_i$  (Loewner case) &  $\nu_j$  and  $\eta_i$  (AAA case)  
As chosen by Trefethen et al. (left) & Random case (right)

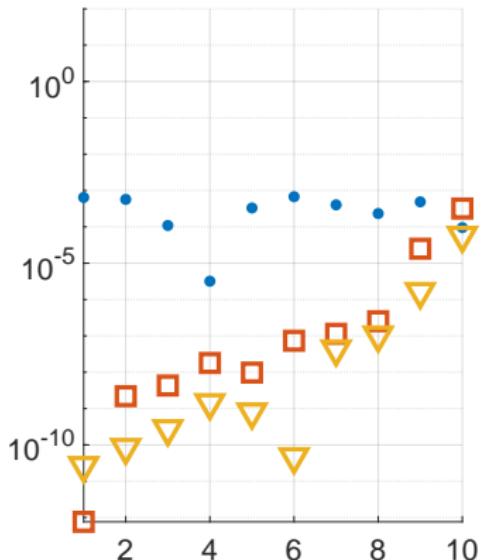


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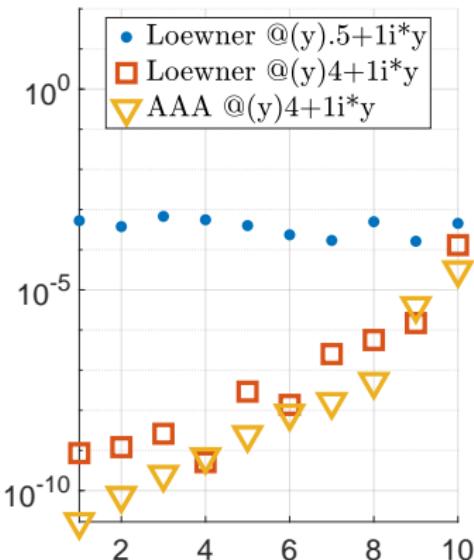
Zeros computation on the rational interpolant

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

Real part mismatch



Imag. part mismatch

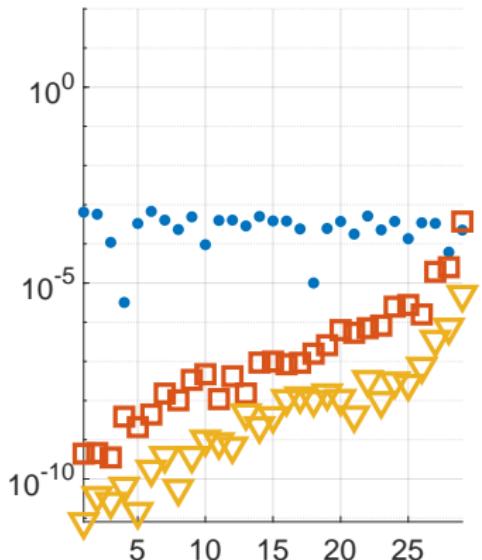


# Numerical illustration

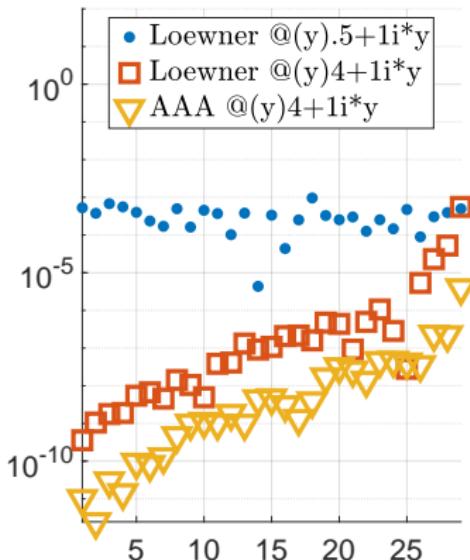
Zeros computation on the rational interpolant

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

Real part mismatch



Imag. part mismatch

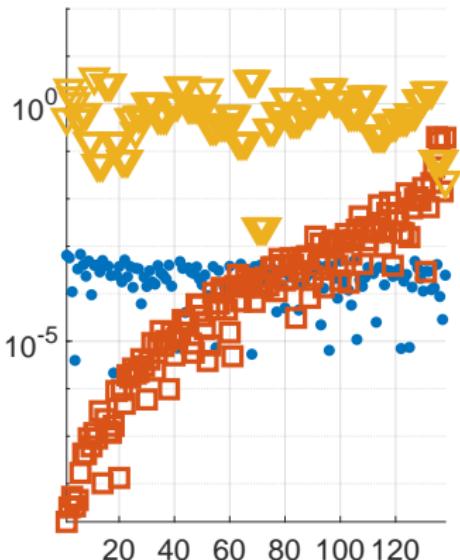


# Numerical illustration

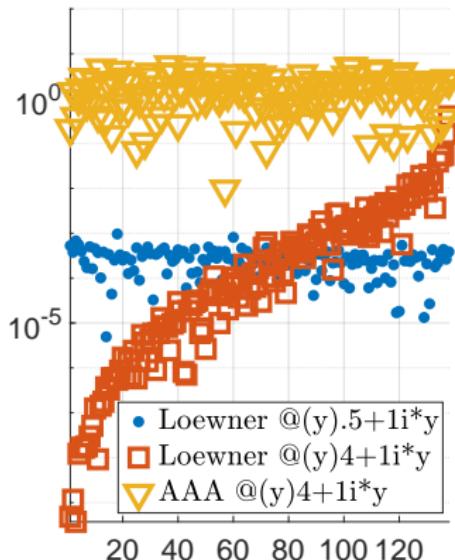
Zeros computation on the rational interpolant

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

Real part mismatch



Imag. part mismatch

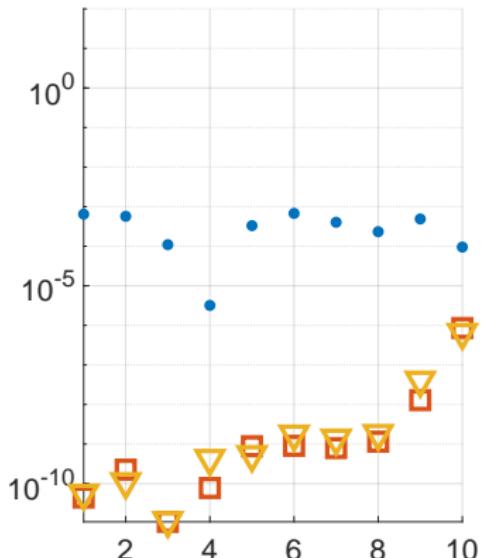


# Numerical illustration

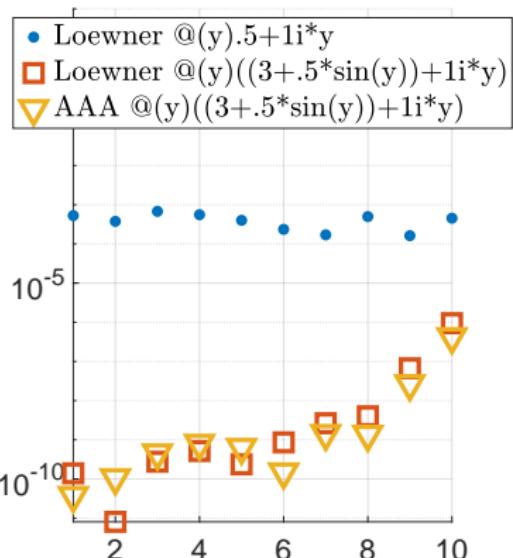
Zeros computation on the rational interpolant

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

Real part mismatch



Imag. part mismatch

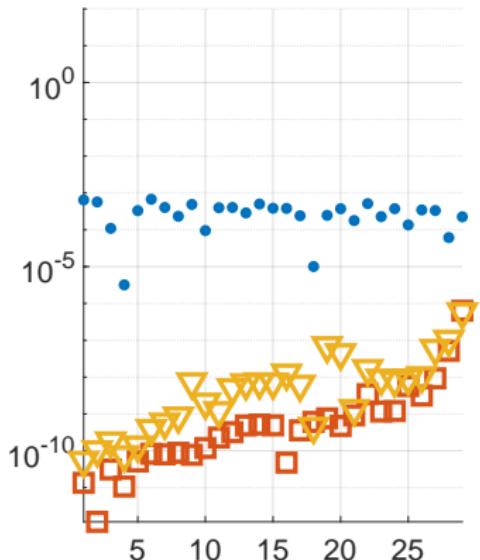


# Numerical illustration

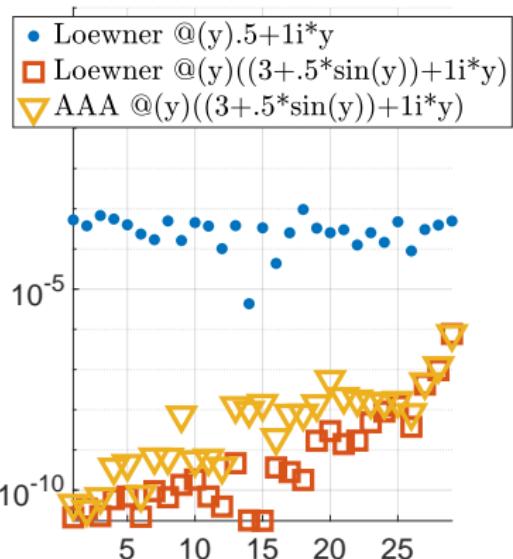
Zeros computation on the rational interpolant

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

Real part mismatch



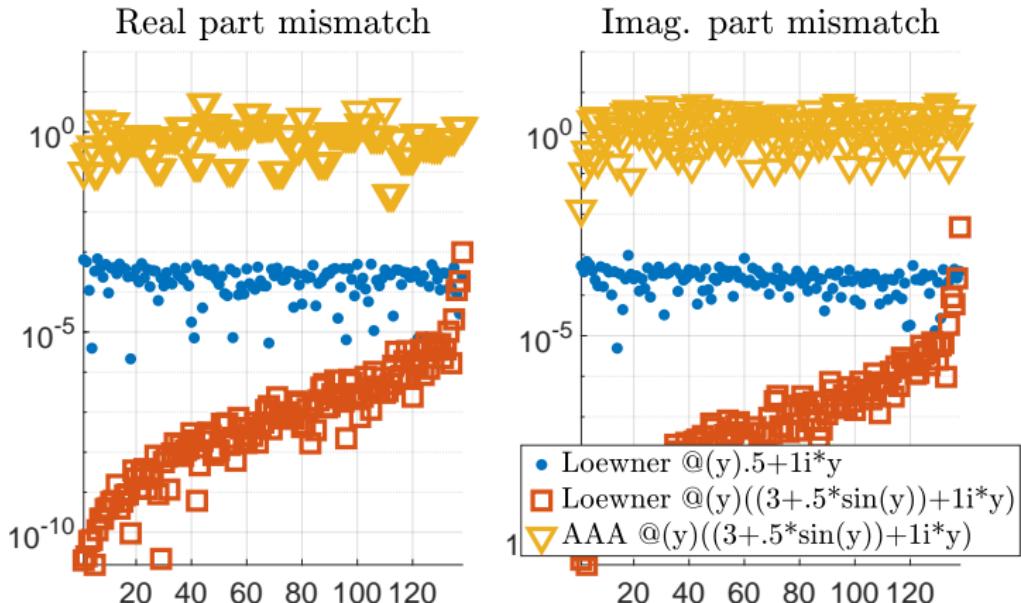
Imag. part mismatch



# Numerical illustration

Zeros computation on the rational interpolant

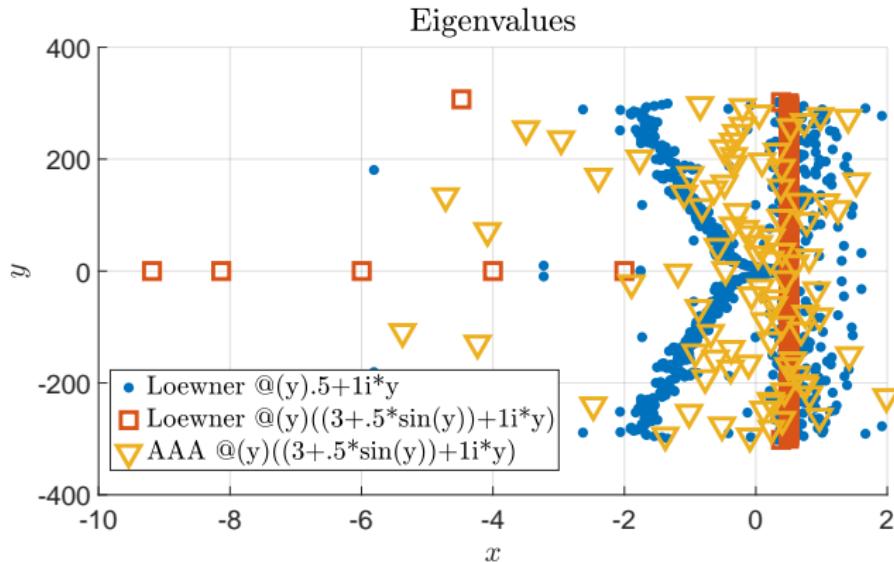
Interval  $[0, 300]$ ,  $n = 1000$ ,  $\zeta(3 + 0.5 \sin(y) + iy)$



# Numerical illustration

Zeros computation on the rational interpolant

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

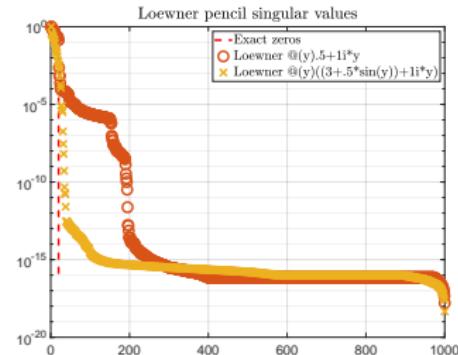
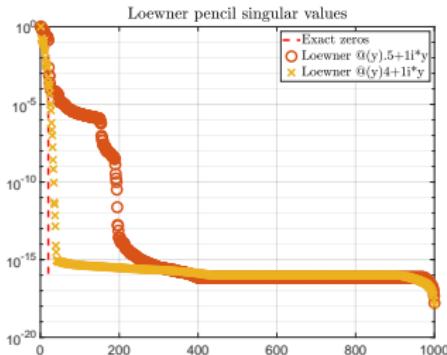


Loewner pencil finds trivial and non-trivial zeros, whatever the choice of interpolation points

# Numerical illustration

## Pencil singular values

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

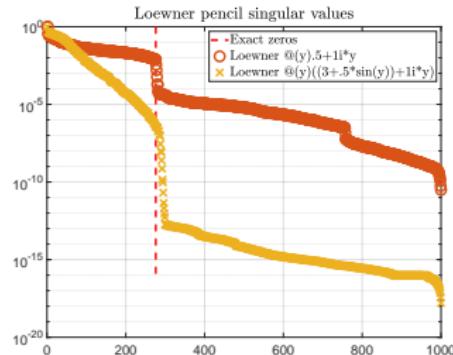
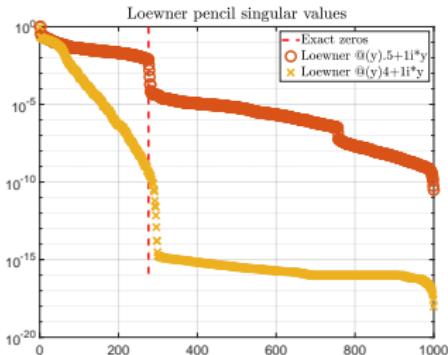


Loewner pencil detects the number of non-trivial zeros

# Numerical illustration

## Pencil singular values

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

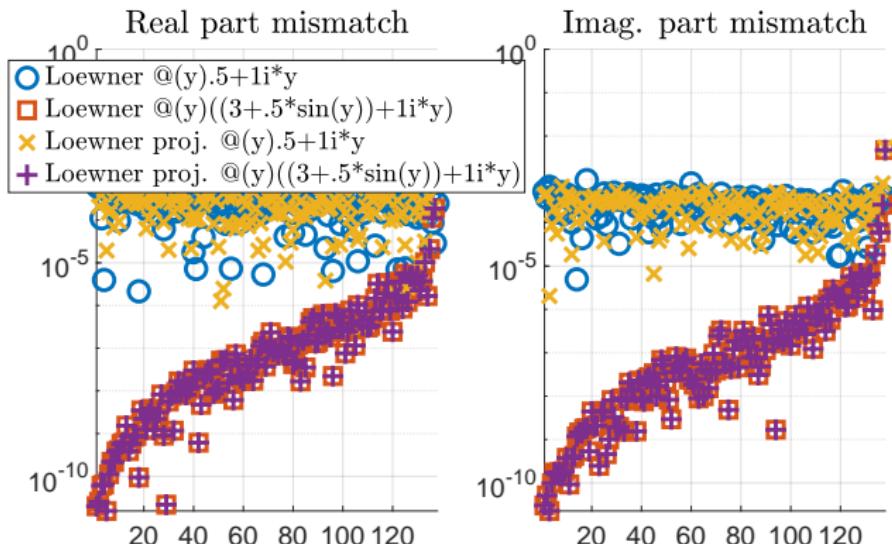


Loewner pencil detects the number of non-trivial zeros

# Numerical illustration

Zeros computation with projected rational model

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



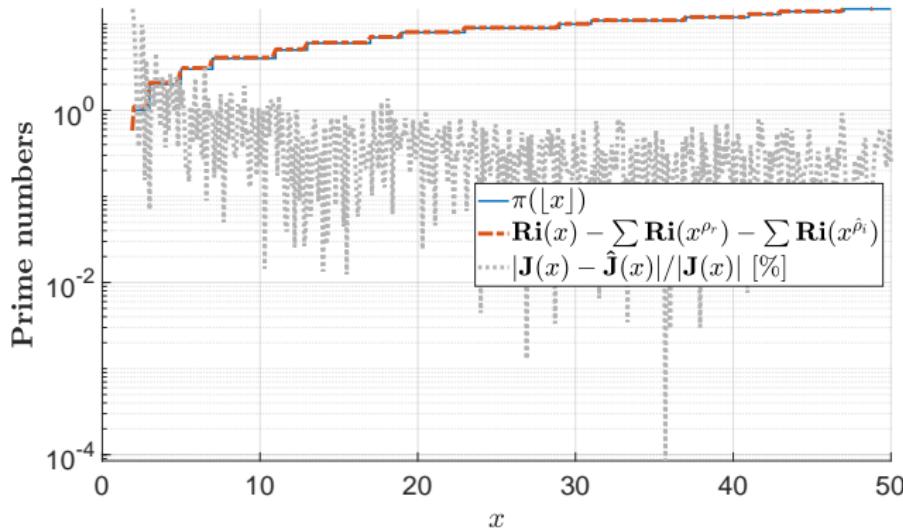
$$\hat{\mathbf{H}}_r(s) = \mathbf{W} \mathbf{X} (-s \mathbf{Y}^T \mathbf{L} \mathbf{X} + \mathbf{Y}^T \mathbf{M} \mathbf{X})^{-1} \mathbf{Y}^T \mathbf{V}$$

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

# Conclusions

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

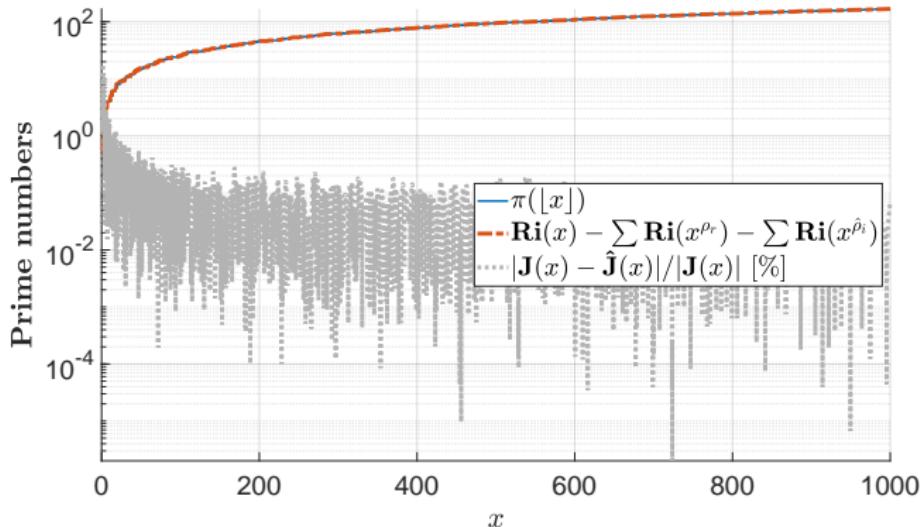
$$\hat{\mathbf{J}}(x) = \mathbf{Ri}(x) - \sum_{\rho_r} \mathbf{Ri}(x^{\rho_r}) - \sum_{\rho_i i=1 \dots 500} \mathbf{Ri}(x^{\hat{\rho}_i})$$



# Conclusions

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

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



# Conclusions

Take home message<sup>2</sup>

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

- ▶ **Loewner** and **AAA** find "accurate" rational models
- ▶ **Loewner pencil** catches the exact number of  $\zeta$  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)



<sup>2</sup>Ack. to David Louapre (<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



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TECHNICAL SYSTEMS  
MAGDEBURG



# Appendix

## Interpolation points?

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

