

# Controlled analytic continuation of Matsubara correlation functions using minimal pole representation

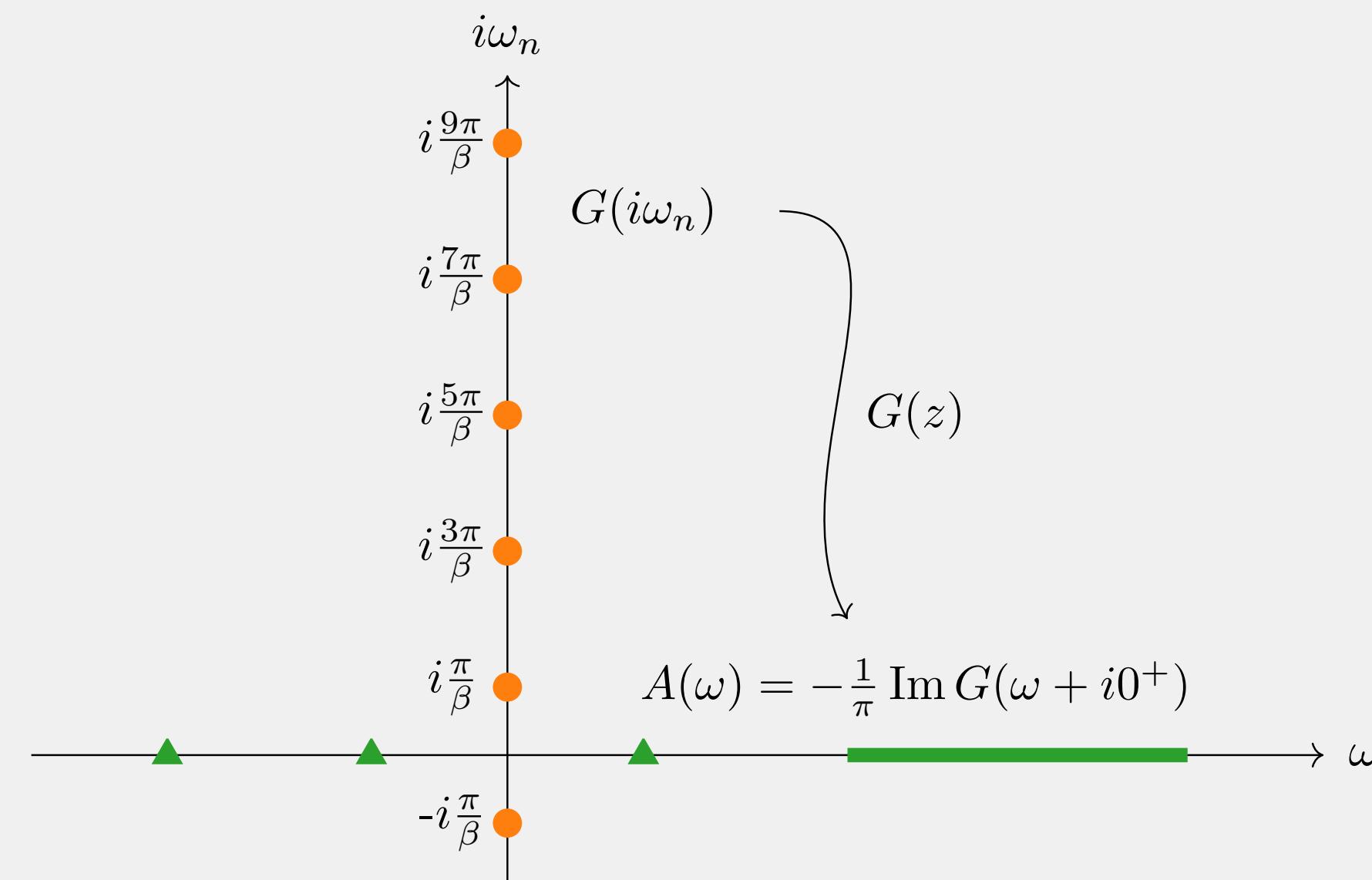


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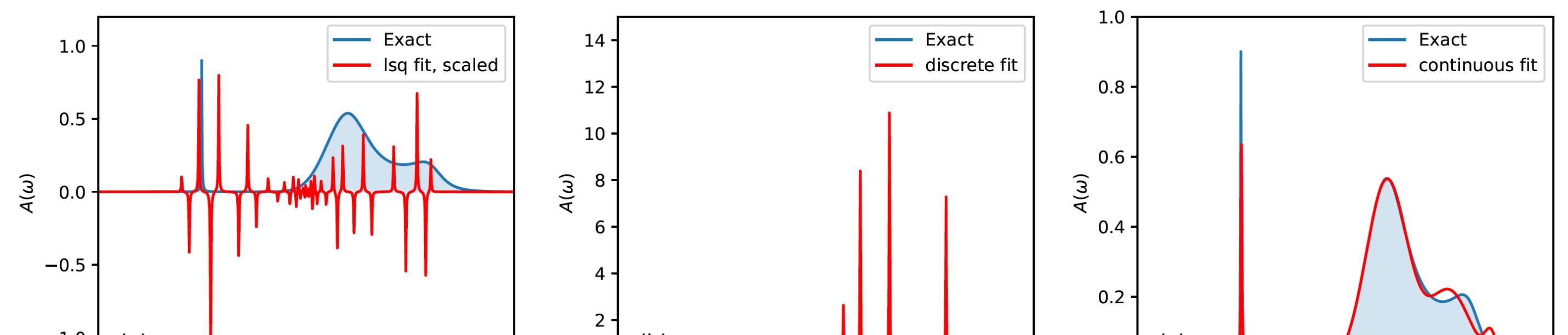
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## Background

- Simulations are performed on the imaginary axis (orange circles)
- Physical interpretation requires evaluation right above singularities (green)
- Bridging the two requires numerical analytic continuation (NAC)



## Key problems



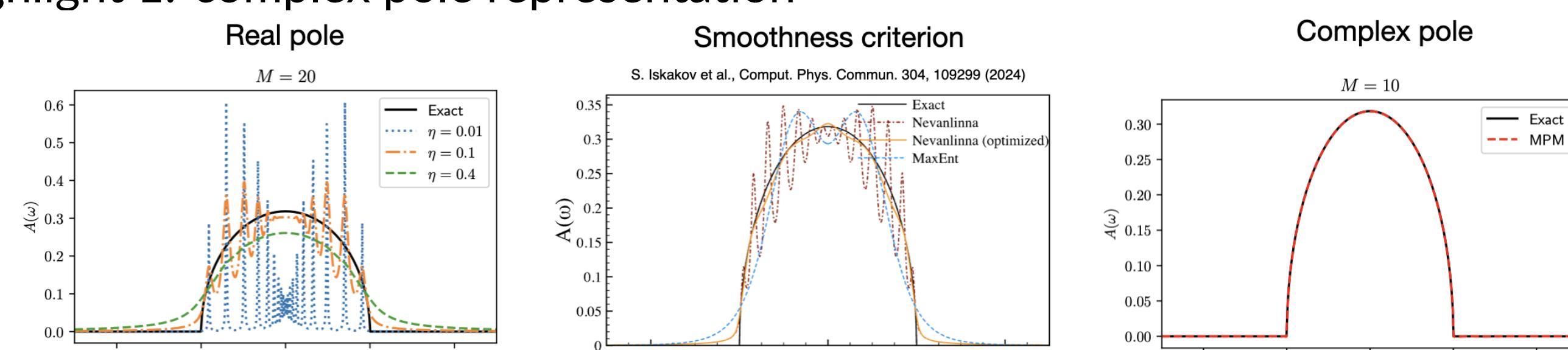
- Infinite solutions  $\{A_i\}$  at any finite precision, even after excluding non-causal ones
- Different solutions differ a lot:  $\sup_{ij} |A_i(\omega) - A_j(\omega)| = +\infty$  for any  $\omega \in (-\infty, +\infty)$
- Without prior knowledge, no clear criterion that which solution should be chosen
- Impossible for *any* method to completely eliminate bias in the recovered solution

## Our Contribution

- We demonstrate that bias can always be reduced as data quality improves
- A general-purpose method applicable to all cases: fermionic / bosonic, noisy / noise-free, diagonal / off-diagonal / matrix-valued, discrete / continuous
- Systematically improvable – the spectral function can, in principle, be recovered to arbitrary precision given sufficiently accurate input data

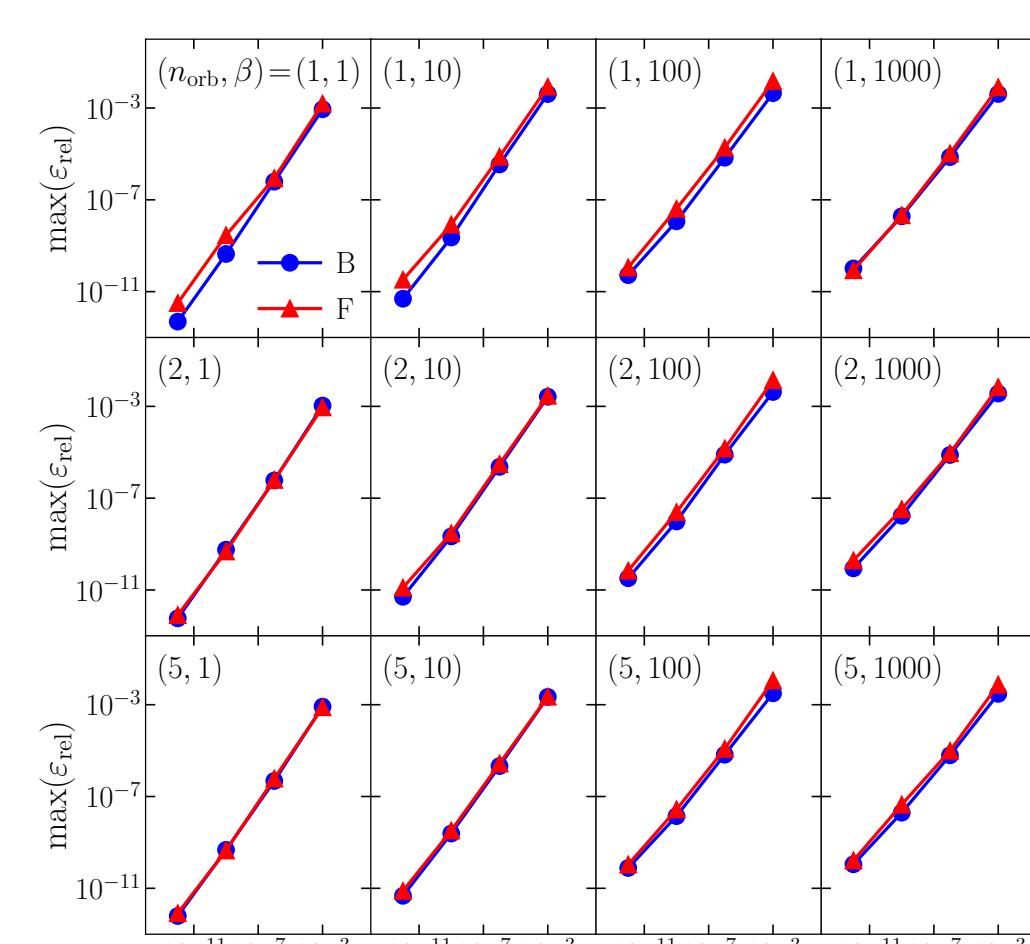
## Highlights

### Highlight 1: complex pole representation



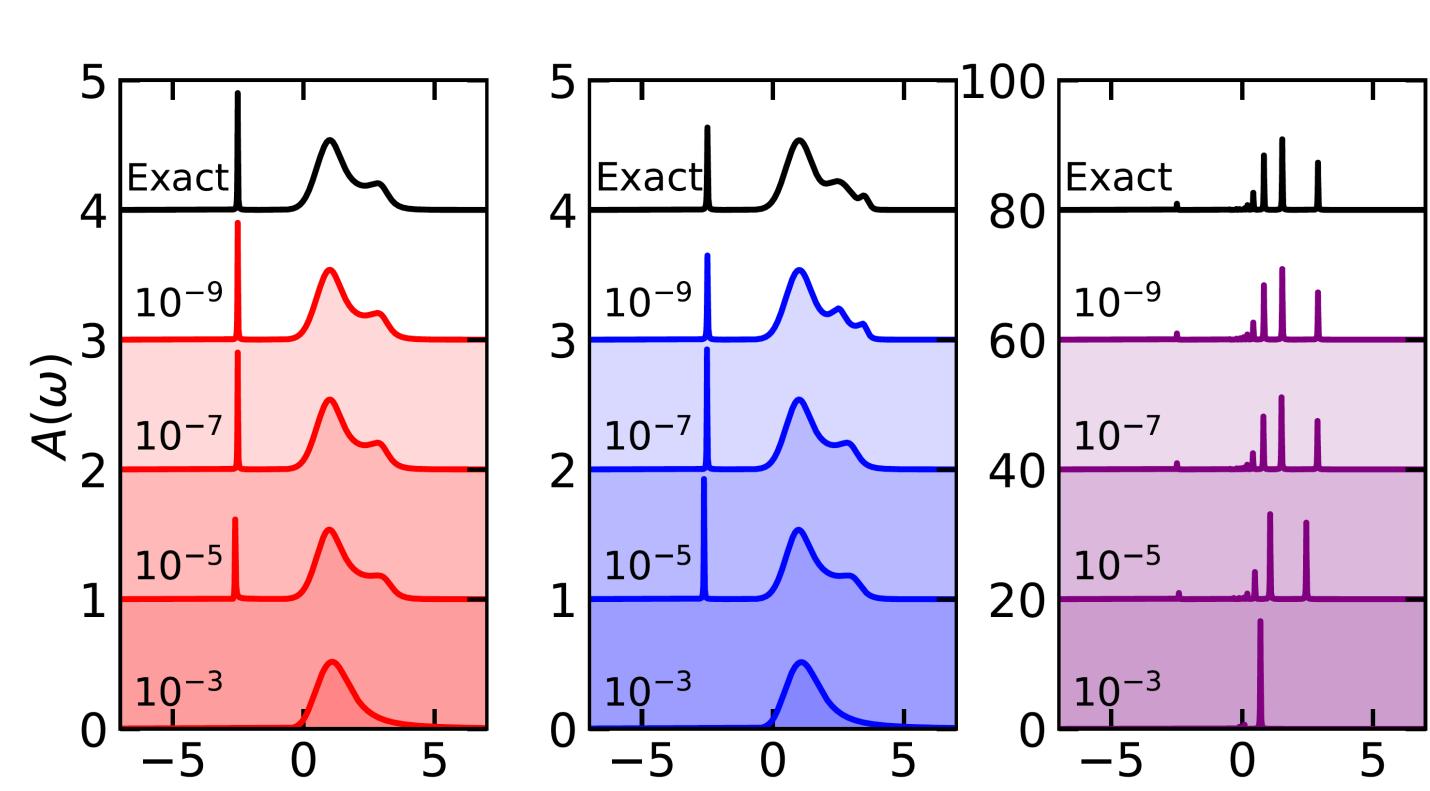
- Real-axis poles capture discrete features, as in conventional approaches
- Branch cuts are approximated by poles in the lower-half complex plane
- Numerically sufficient to capture all features of the spectrum [3]

### Highlight 2: controlled approximation for $G(iy)$



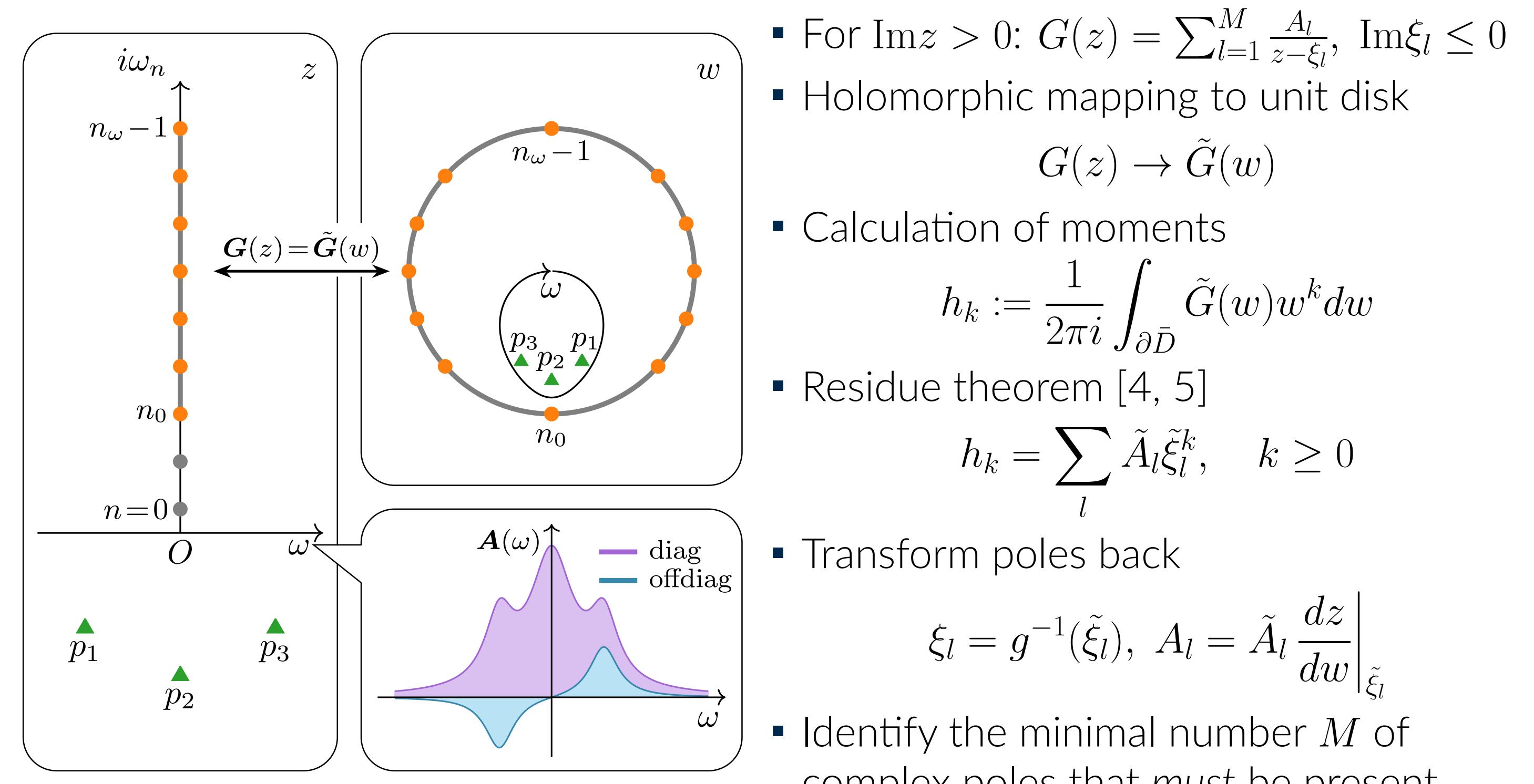
- Fit  $G(i\omega_n)$  with the minimal # of exponentials
- Suppress spurious oscillations in between
- Vary  $L$  in ESPRIT to identify oversampling points

### Highlight 3: minimal information principle



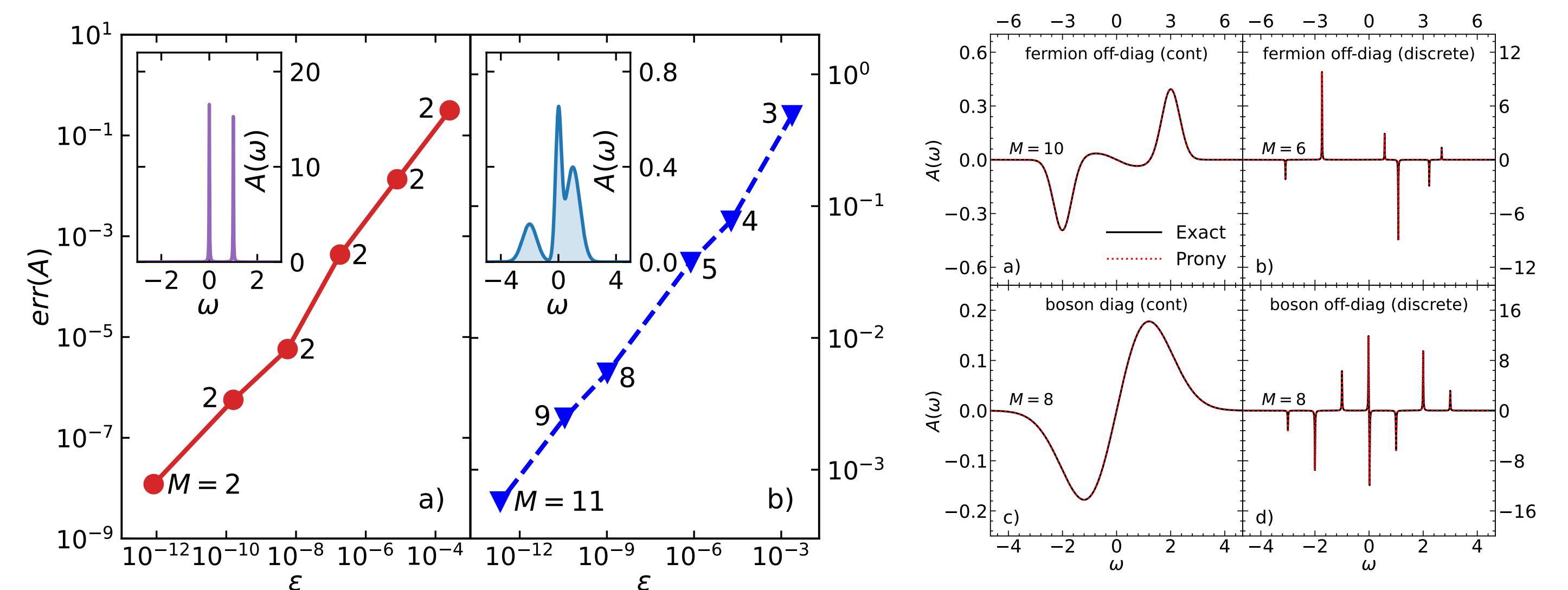
- Indistinguishable under large noise
- Distinguishable as noise decreases
- No artificial features are recovered
- Symmetry constraints accelerate convergence for discrete systems

## Method

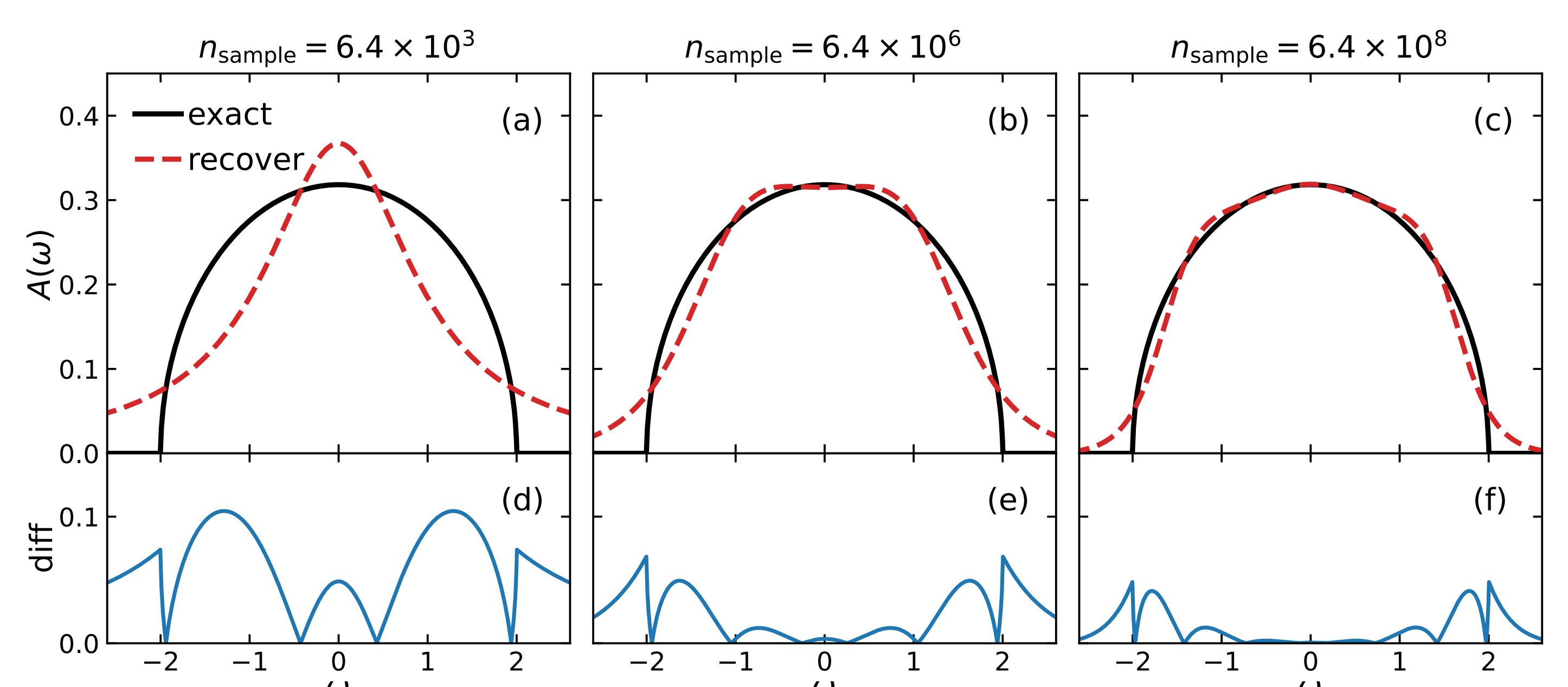


- For  $\text{Im } z > 0$ :  $G(z) = \sum_{l=1}^M \frac{A_l}{z - \xi_l}$ ,  $\text{Im } \xi_l \leq 0$
- Holomorphic mapping to unit disk  $G(z) \rightarrow \tilde{G}(w)$
- Calculation of moments  $h_k := \frac{1}{2\pi i} \int_{\partial\bar{D}} \tilde{G}(w) w^k dw$
- Residue theorem [4, 5]  $h_k = \sum_l \tilde{A}_l \tilde{\xi}_l^k$ ,  $k \geq 0$
- Transform poles back  $\xi_l = g^{-1}(\tilde{\xi}_l)$ ,  $A_l = \tilde{A}_l \left. \frac{dz}{dw} \right|_{\tilde{\xi}_l}$
- Identify the minimal number  $M$  of complex poles that must be present

## Results



- $\epsilon$ : precision;  $\text{err}(A) = \int_{\mathbb{R}} d\omega |A - A_{\text{cont}}|$
- Systematically improvable, independent of  $\beta$  and system specifics
- Robust to  $n_\omega$ :  $(n_\omega)_{\min} \gtrsim 2n_{\text{pole}} + 1$
- Noiseless, double precision
- Spectral function can be accurately recovered
- Versatile across different systems



- CT-HYB: the recovered results improve progressively as runtime increases
- Maxent: unable to resolve distinct bands
- Nevanlinna/Carathéodory: state-of-the-art, uses multiprecision arithmetic, not robust to noise, introduces artificial broadening, requires 800 CPU hrs
- MPM: operates in double precision, robust to noise, no artificial broadening, yields analytic expressions, 80 CPU hrs (MiniPole) and 160 CPU secs (MiniPoleDLR)



- Code available at: <https://github.com/Green-Phys/MiniPole>
- Also available via PyPI (`pip install mini_pole`) and Zenodo

Openings at the University of Warsaw in Poland as part of ERC Advanced grant Quantum Algorithms. Visit [gull-group.org](http://gull-group.org) and apply to [emanuel.gull@gmail.com](mailto:emanuel.gull@gmail.com).

## References

- [1] L. Zhang and E. Gull, Phys. Rev. B 110, 035154 (2024).  
[2] L. Zhang, Y. Yu, and E. Gull, Phys. Rev. B 110, 235131 (2024).  
[3] L. Zhang, A. Erpenbeck, Y. Yu and E. Gull, arXiv:2504.01163.  
[4] L. Ying, J. Comput. Phys. 469, 111549 (2022).  
[5] L. Ying, J. Sci. Comput. 92, 107 (2022).