

A Contour Integral-based Parallel Eigensolver for Nonlinear Eigenvalue Problems

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Introduction

- Nonlinear Eigenvalue Problem (NEP)

$$T(\lambda)x = \mathbf{0},$$

where $T(\lambda)$ is a matrix-valued function.

- We find eigenvalues in a given domain on the complex plane
 - Interior eigenvalue problem
 - Large-scale sparse matrices

Large and sparse NEP



Projection

Small size NEP

Introduction

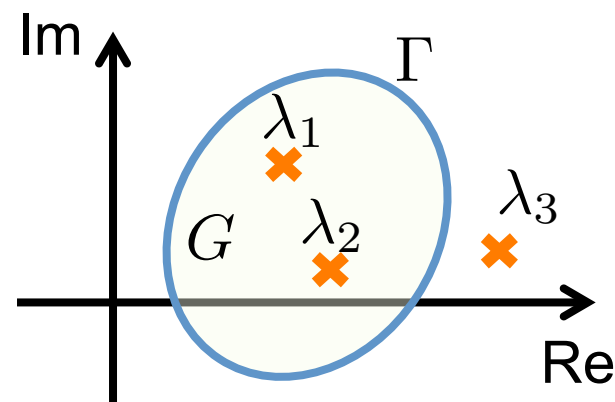
■ Contour Integral-based Methods

➤ GEPs

- SS-Hankel (S and Sugiura, 2003)
- SS-RR (S and Tadano, 2007)
- FEAST (Polizzi, 2009)
- SS-Arnoldi (Imakura, Du and S, 2013)

➤ NEPs

- SS-Hankel (Asakura, S, et al. 2009)
- Beyn's method (Beyn, 2012)
- SS-RR (Yokota and S, 2013)



Map of CI-based eigensolvers

(under investigation)

Target eigenproblem
in Γ

$$T(\lambda_i)\mathbf{x}_i = \mathbf{0}$$

Rayleigh-Ritz procedure
with filtered subspace

$$V^H T(\theta_i) V \mathbf{y}_i = \mathbf{0}$$

SS-RR (GEP: 2007, NEP: 2013)

FEAST (GEP: 2009)

**(Implicit)
Linear EP**

Corresponding linear
eigenproblem

$$A\mathbf{x}_i = \lambda_i \mathbf{x}_i$$

Petrov-Galerkin condition
with filtered subspace

$$W^H A V \mathbf{y}_i = \theta_i W^H V \mathbf{y}_i$$

SS-Hankel (GEP:2003, NEP:2009)

Rayleigh-Ritz procedure
with filtered subspace

$$V^H A V \mathbf{y}_i = \theta_i \mathbf{y}_i$$

Beyn's method (NEP:2012)

Arnoldi method
with filtered subspace

$$V^H A V \mathbf{y}_i = \theta_i \mathbf{y}_i$$

SS-Arnoldi (GEP:2013)

**Improvement is
now under investigation**

Spectral Projection using Contour Integral

Contour Integral-based Eigensolver

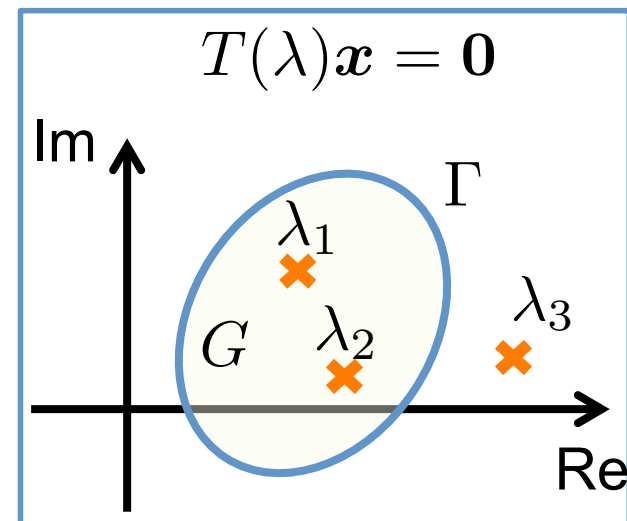
- Contour integral for a rational function

$$\frac{1}{2\pi i} \oint_{\Gamma} \sum_{i=1}^n \frac{\nu_i}{z - \lambda_i} dz = \sum_{\lambda_i \in G} \nu_i$$

- Spectral decomposition of $T^{-1}(z)$
(Keldish 1951, Beyn 2012)

$$T^{-1}(z) = \sum_{i=1}^k \frac{P_i}{z - \lambda_i} + R(z)$$

λ_i : eigenvalue, P_i : spectral projection with respect to λ_i , k : #eigs in Γ
(for simplicity, we consider the case that λ_i is simple)



Localization of spectral projection using contour integral

$$P_{\Gamma} = \frac{1}{2\pi i} \oint_{\Gamma} T^{-1}(z) dz = \sum_{\lambda_i \in G} P_i$$

Numerical Quadrature

- Contour integral is approximated by numerical quadrature

$$\sum_{\lambda_i \in G} P_i = \frac{1}{2\pi i} \oint_{\Gamma} T^{-1}(z) dz \approx \sum_{j=1}^N w_j T^{-1}(z_j), \quad (N \gg 1)$$

z_j : quadrature point, w_j : quadrature weight

- Apply for an arbitrary vector \mathbf{v}

$$\sum_{\lambda_i \in G} P_i \mathbf{v} \approx \sum_{j=1}^N w_j T^{-1}(z_j) \mathbf{v}$$



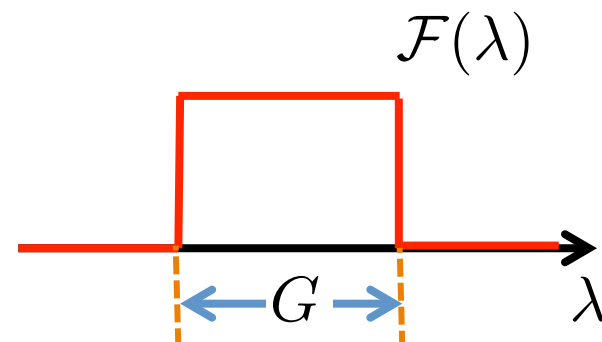
Systems of linear equations at z_j

$$T(z_j) \mathbf{y}_j = \mathbf{v}, \quad j = 1, \dots, N$$

Filter for a Subspace

- Define a filter function

$$\mathcal{F}(\lambda) = \frac{1}{2\pi i} \oint_{\Gamma} \frac{1}{z - \lambda} dz$$

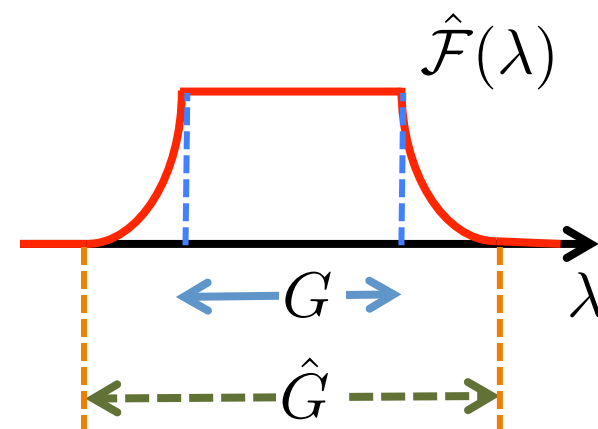


- A projection is represented by

$$\frac{1}{2\pi i} \oint_{\Gamma} T^{-1}(z) dz = \sum_i \mathcal{F}(\lambda_i) P_i + \tilde{R}(z)$$

- For numerical quadrature, the filter function is given by

$$\hat{\mathcal{F}}(\lambda) = \sum_{j=1}^N \frac{w_j}{z_j - \lambda}$$



By taking larger size of subspace, contamination is avoided.

(Ikegam, S and Nagashima, 2010)

Quadrature Rule and Filter

■ Trapezoidal Rule: Ellipse

➤ Quadrature points

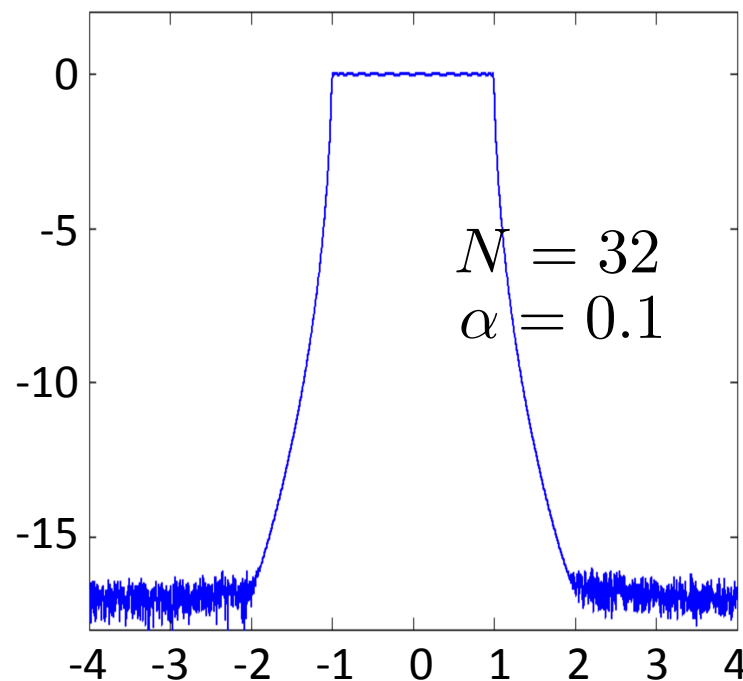
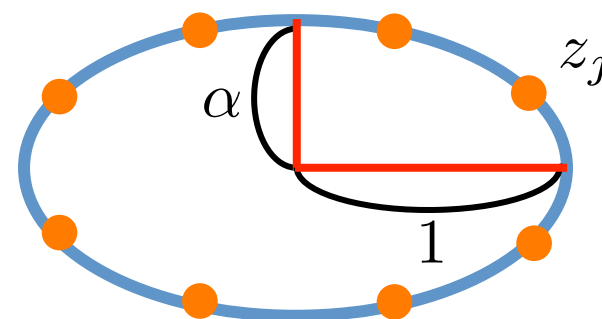
$$z_j = \cos \theta_j + i\alpha \sin \theta_j$$

➤ Quadrature weights

$$w_j = \alpha \cos \theta_j + i \sin \theta_j$$

➤ Filter function

$$\hat{\mathcal{F}}(\lambda) = \sum_{j=1}^N \frac{w_j}{z_j - \lambda}$$

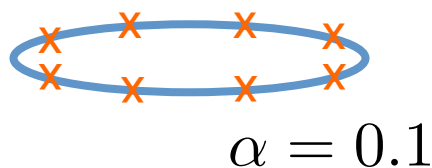
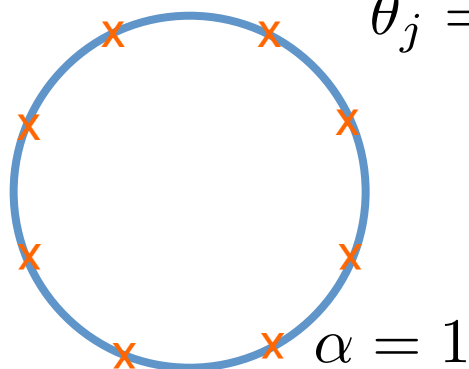


Filter for a Subspace

■ Selection of α

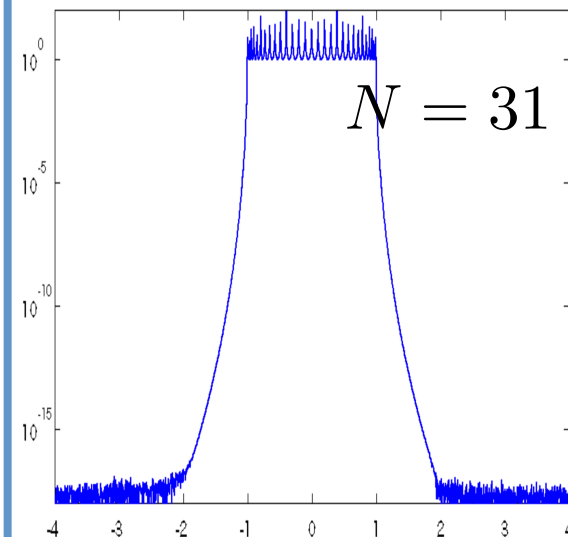
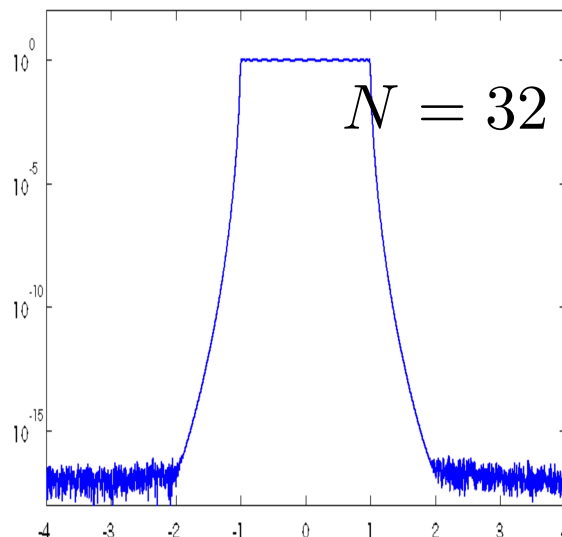
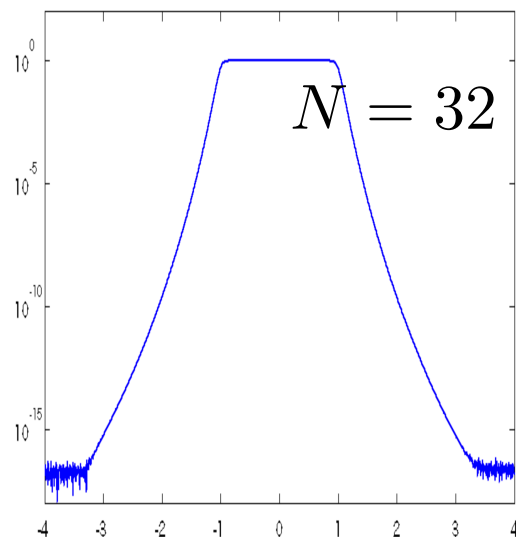
$$z_j = \cos \theta_j + i\alpha \sin \theta_j$$

$$\theta_j = \frac{2\pi}{N} \left(j - \frac{1}{2} \right)$$



$$\theta_j = \frac{2\pi}{N} \left(j - \frac{1}{4} \right)$$

Equivalent to rational interpolation with Chebyshev points (Austin and Trefethen 2013)



SS-Hankel

- Consider a scalar function

$$f(z) := \boldsymbol{w}^H T^{-1}(z) \boldsymbol{v}$$

where \boldsymbol{w} and \boldsymbol{v} are nonzero vectors.

- Apply a root-finding method with contour integral
(cf. Kravanja, S and Van Barel, 1999)
- The problem is reduced to the GEP with Hankel matrices.

$$\mu_k = \frac{1}{2\pi i} \oint_{\Gamma} z^k f(z) dz, \quad k = 0, 1, \dots, 2m - 1$$

$$H_m = (\mu_{i+j-2})_{i,j=1,m} \quad H_m^< = (\mu_{i+j-1})_{i,j=1,m}$$



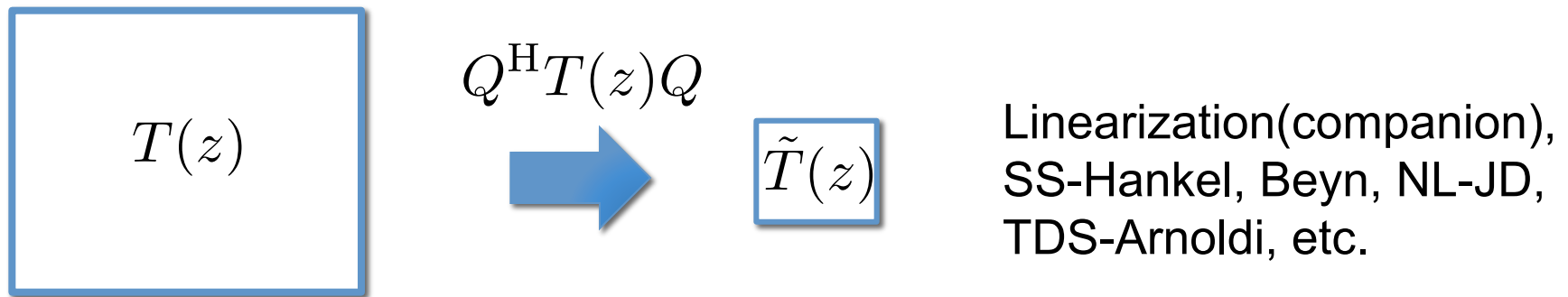
$$H_m^< \boldsymbol{y} = \lambda H_m \boldsymbol{y}$$

■ Filter by Contour Integral:

$$S_k = \frac{1}{2\pi i} \oint_{\Gamma} z^k T^{-1}(z) dz V, \quad k = 0, 1, \dots, M-1$$

➤ Rayleigh-Ritz procedure:

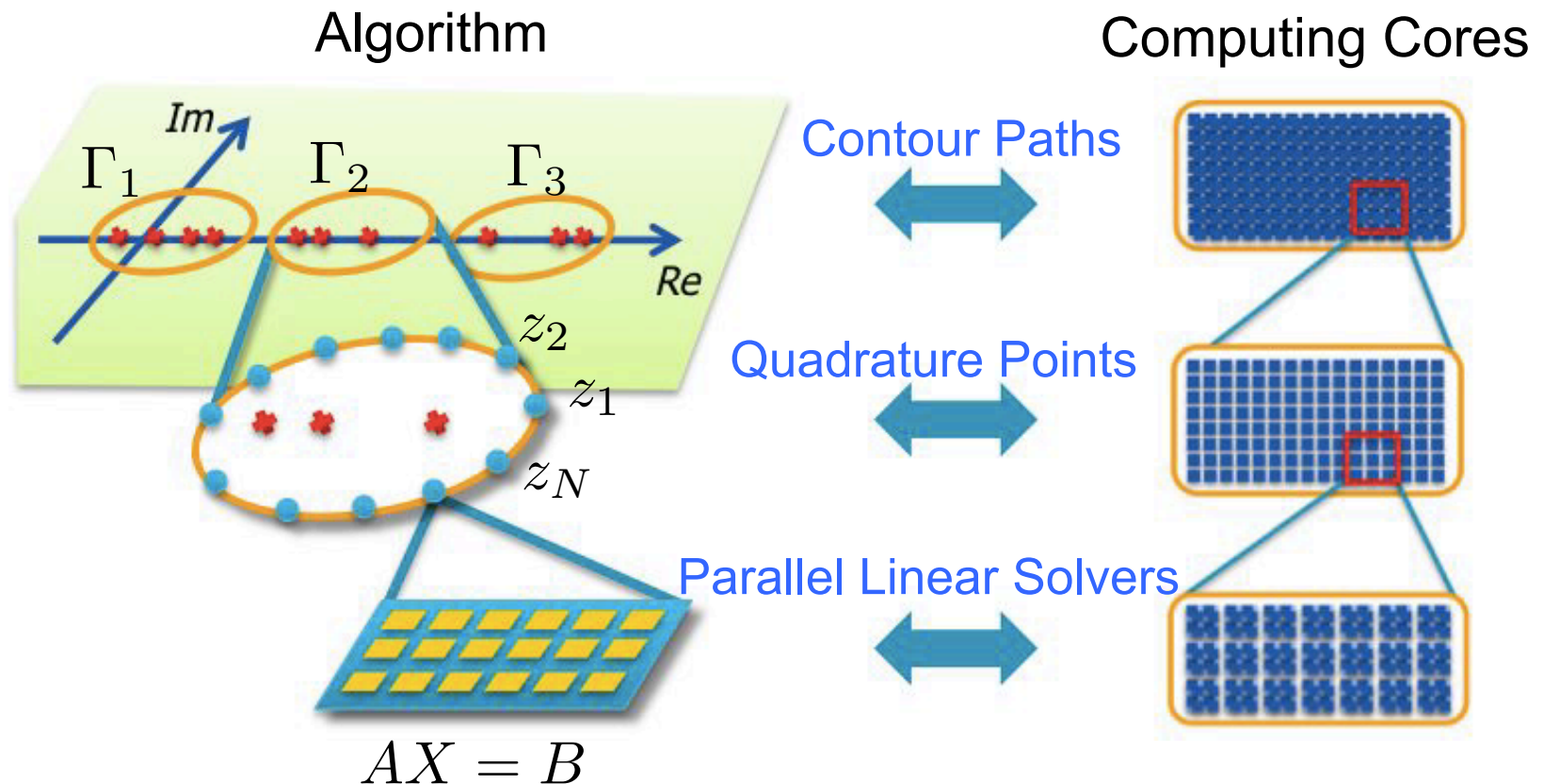
$$S = [S_0, S_1, \dots, S_{M-1}] \quad S = Q \Sigma U^H$$



The projected problem is solved by an appropriate nonlinear eigensolver

Parallel Implementation

- Computing cores are assigned according to a hierarchical structure of the algorithm



Stochastic Estimation of Eigenvalue Count

Stochastic Estimation of Eigenvalue Count

- To find good parameters in the method:

- Number of eigenvalues in Γ is given by

$$m = \frac{1}{2\pi i} \oint_{\Gamma} \text{tr}(T^{-1}(z)T'(z))dz$$

- Approximate the trace of inverse matrix by

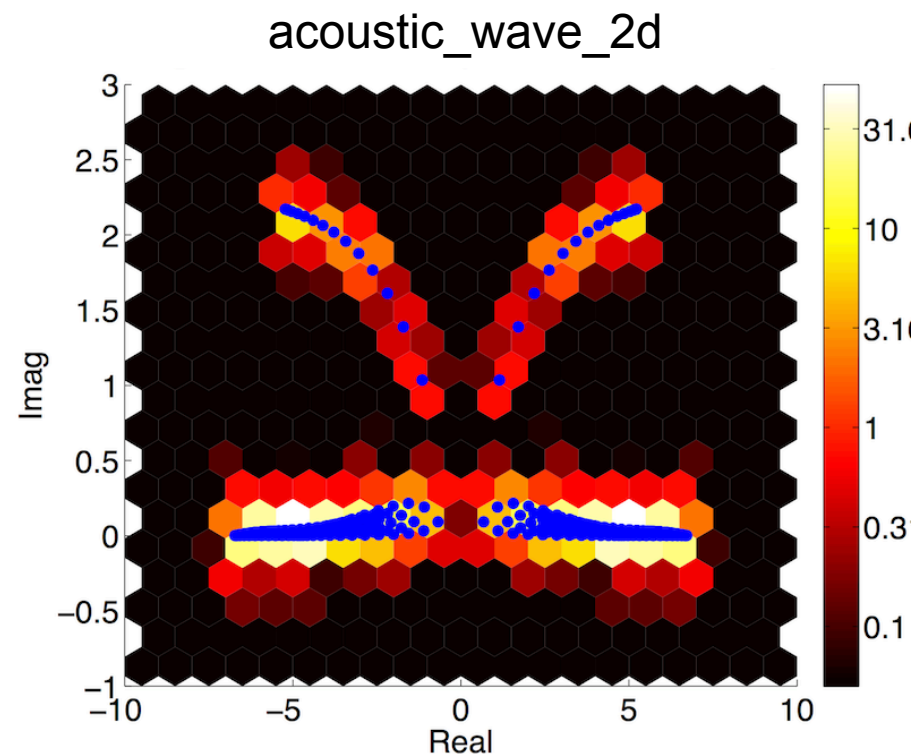
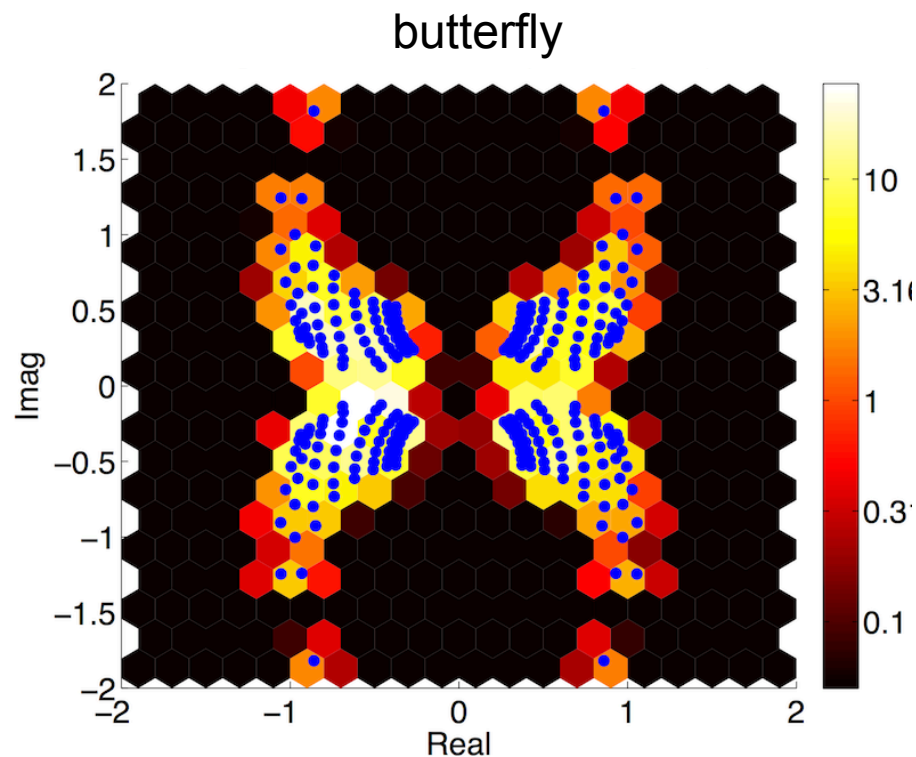
$$\text{tr}(T^{-1}(z)T'(z)) \approx \frac{1}{L} \sum_{\ell=1}^L \mathbf{v}_{\ell}^T T^{-1}(z)T'(z) \mathbf{v}_{\ell}$$

where \mathbf{v}_{ℓ} is a vector of which elements are taken as 1 or -1 with equal probability.

GEP: Futamura, Tadano and S (2010), NEP: Maeda, Futamura and S (2011)

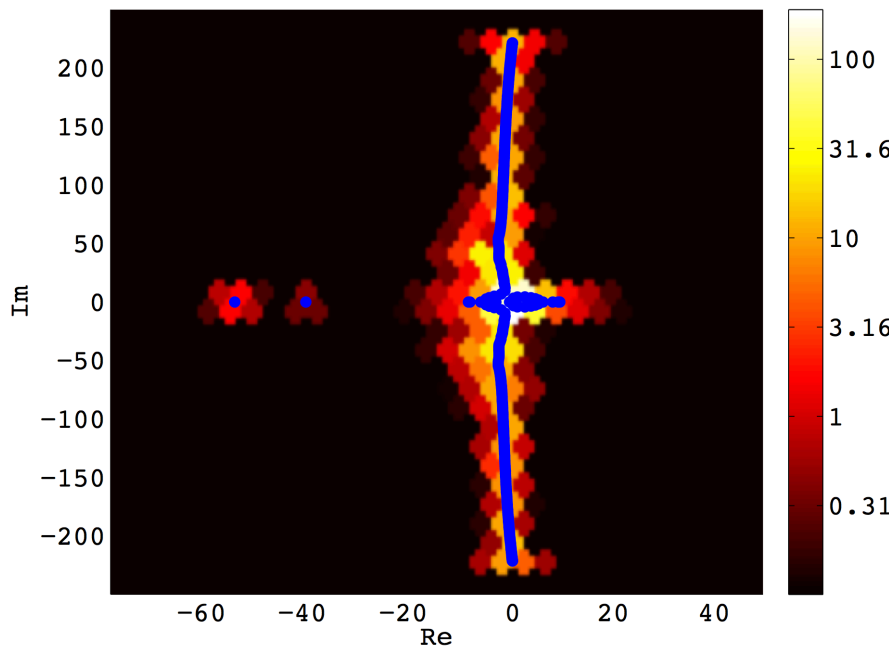
Stochastic Estimation of Eigenvalue Counts

- Find areas that include eigenvalues
 - Estimate eigenvalue density
 - Put many circles on the complex plane
 - Examples form NLEVP

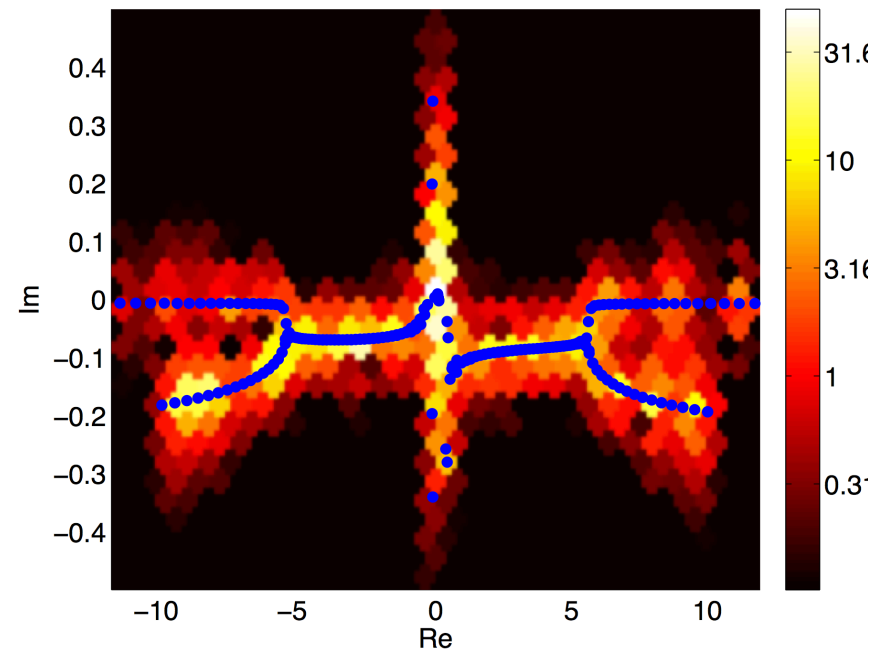


Stochastic Estimation of Eigenvalue Counts

planar_waveguide



plasma_drift



Numerical Examples

Numerical Example

- Test problem: NLEVP: plasma_drift

- Cubic PEP, Matrix dim. = 512

- Solver: SS-RR + polyeig

- Domain: center = 1.5, radius = 0.8

Parameters:

N = 32 (#quad points)

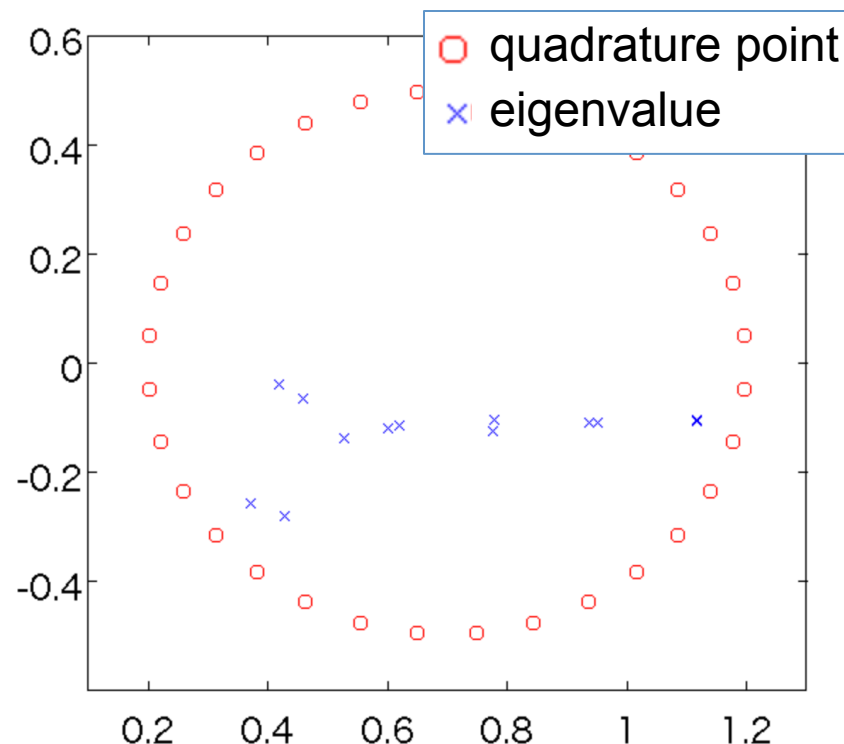
M = 8 (moment deg.)

L = 16 (block size)

- 19 eigs are found

- Subspace size = 47

- max(res) = $1.8e-12$



Numerical Example

- Test problem: NLEVP: orr_sommerfeld

- Quartic PEP, Matrix dim. = 5,000

- Solver: SS-RR + polyeig

- Domain: center = $2 - i$, radius = 1

Parameters:

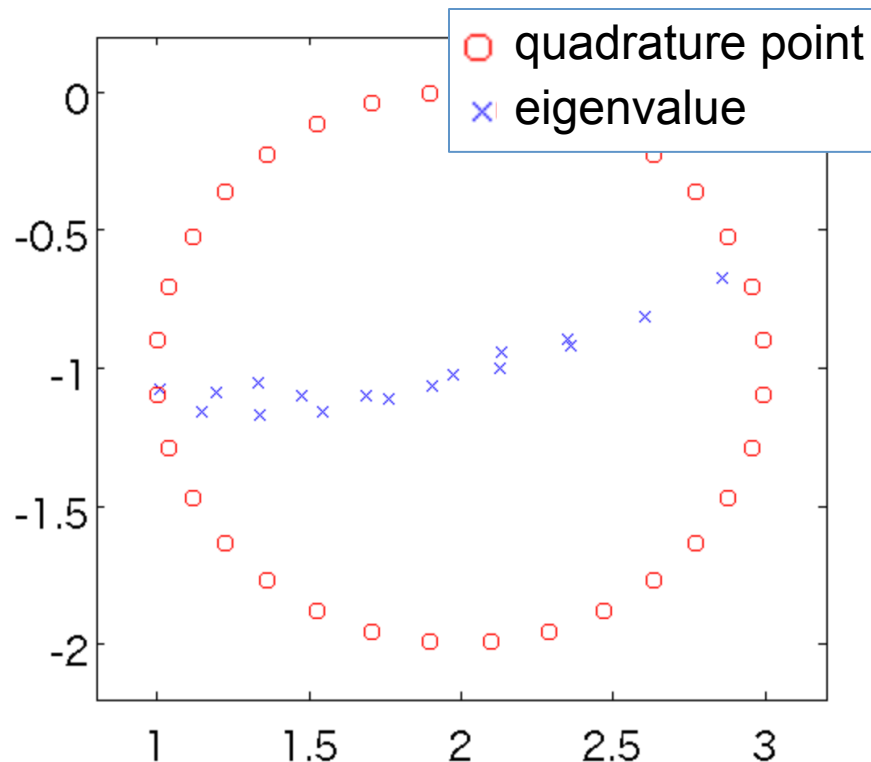
$N = 32$ (#quad points)

$M = 8$ (moment deg.)

$L = 16$ (block size)

- 17 eigs are found

- Subspace size = 49
- $\max(\text{res}) = 1.0\text{e-}15$



Numerical Example

- Test problem: NLEVP: butterfly

- Quartic PEP, Matrix dim. = 10,000

- Solver: SS-RR + polyeig

- Domain: center = $0.253 + 0.236i$, radius = 0.005

Parameters:

$N = 32$ (#quad points)

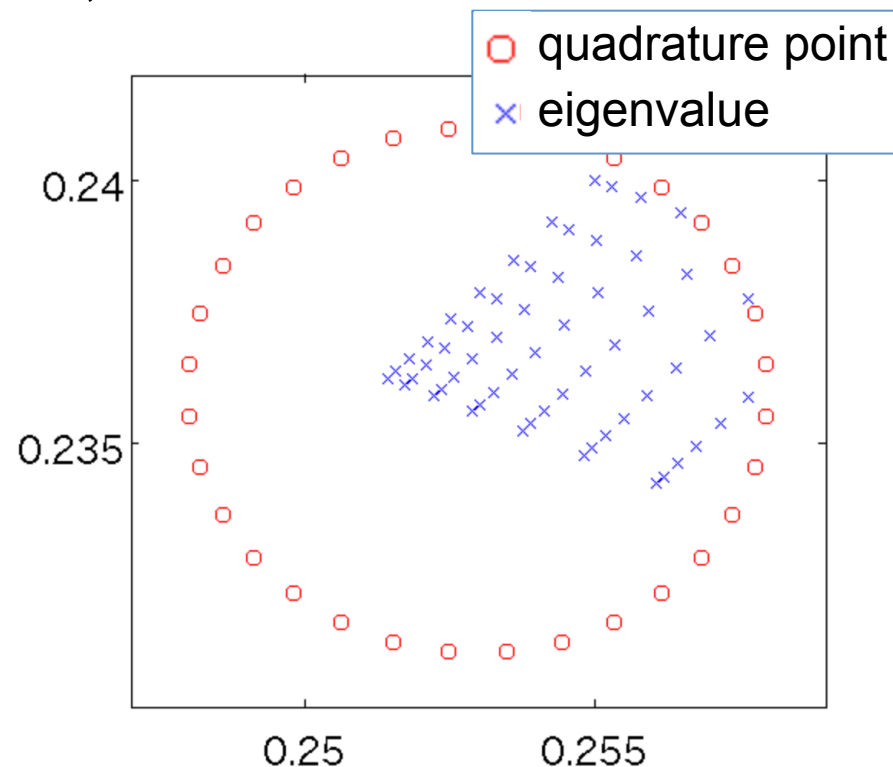
$M = 8$ (moment deg.)

$L = 32$ (block size)

- 58 eigs are found

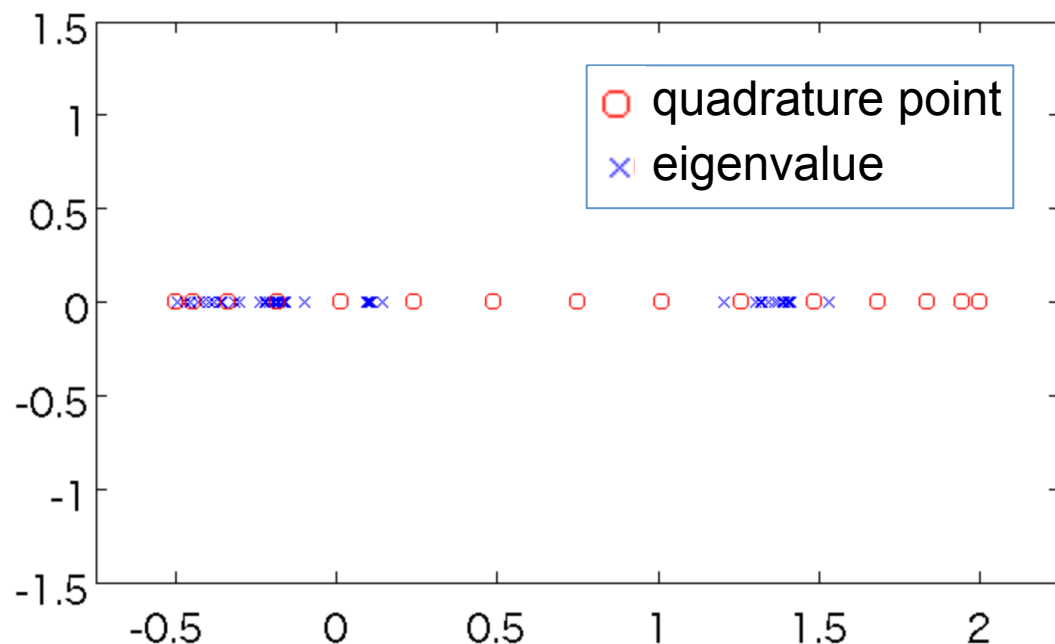
- Subspace size = 74

- $\max(\text{res}) = 2.4e-12$



Numerical Example

- Test problem: NLEVP: schrodinger
 - QEP, Matrix dim. = 1,998
- Solver: SS-RR + polyeig
 - Interval: $[-0.5, 2]$ (real quadrature points)
Parameters:
N = 15 (#quad points)
M = 4 (moment deg.)
L = 32 (block size)
 - 58 eigs are found
 - Subspace size = 96
 - $\max(\text{res}) = 1.0\text{e-}12$



Numerical Example

- Test problem:
Simulation of the international linear collider

$$T(z) = K - z^2 M + i \sum_{j=1}^t \sqrt{z^2 - \sigma_j^2} W_j,$$

where $t = 1$, $\sigma_1 = 0$.

- Test environment: Cray-XT4 at NERSC @Berkeley
- Compiler: PGI Fortran
- Linear solver: SuperLU_DIST
- Eigensolver: SS-Hankel

Numerical Example 3 (NEP)

- Strong scalability for nonlinear (NEP) case
 - Matrix size: 2,738,556
- Two contour paths are used.
 - The number of quadrature points is $N = 32$.
 - 64 linear systems are solved in total.

#cores	256	512	1024	2048
time(sec.)	2513	1273	661	334
speedup	-	1.97	1.93	1.92

(Yamazaki et al., 2013)

Summary

- SS method: Contour integral-based method
 - Applicable for various problems, SEPs, GEPs and NEPs
 - SS-RR: Large sparse NEP is projected to a small size NEP
- Software
 - Fortran95+MPI: z-Pares (GEPs)
 - C: CISS in SLEPc (GEPs)
 - MATLAB: sseig_gep, sseig_pep, sseig_nep
- Future Work
 - Development of high performance software
 - Extension to NEPs
 - Prediction of appropriate parameters