

A parallel eigensolver using contour integration for generalized eigenvalue problems in molecular simulation

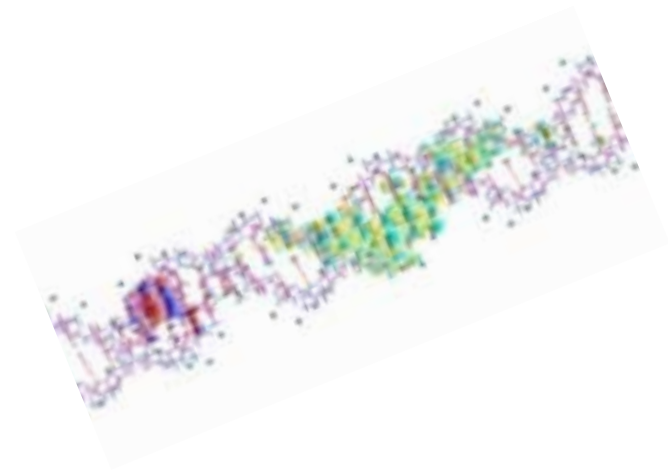
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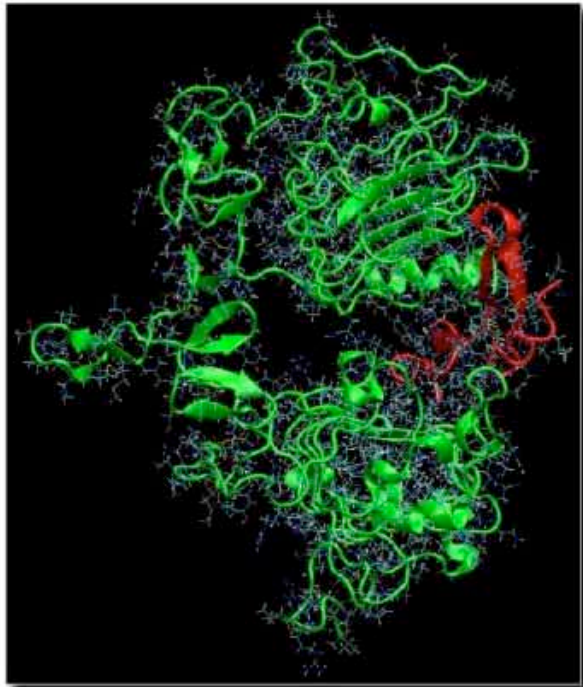
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- Conclusions



Molecular Orbital Computation

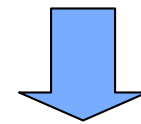
Design of Anticancer Drugs



EGFR
(Epidermal Growth Factor Receptor)

Schrödinger Equation

$$H\Psi = E\Psi$$

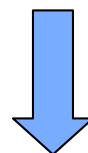


Hartree-Fock
approximation

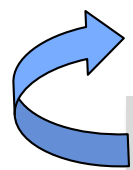
Generalized Eigenvalue Problems

Matrix Generation

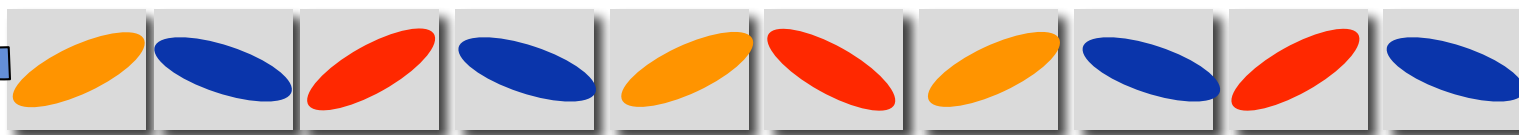
A large molecule is separated to small segments.



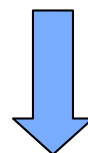
FMO (Fragment MO) method



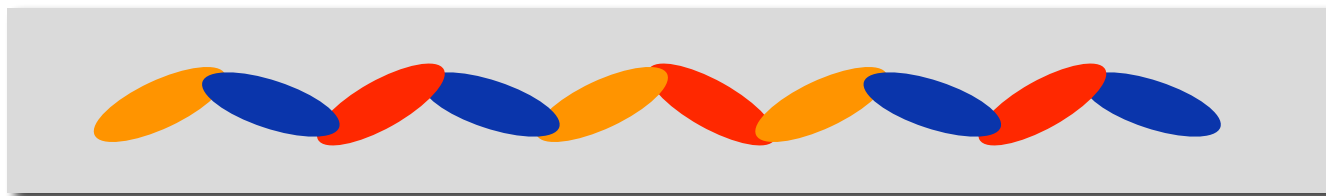
SCF for each segment



(small eigenproblems)



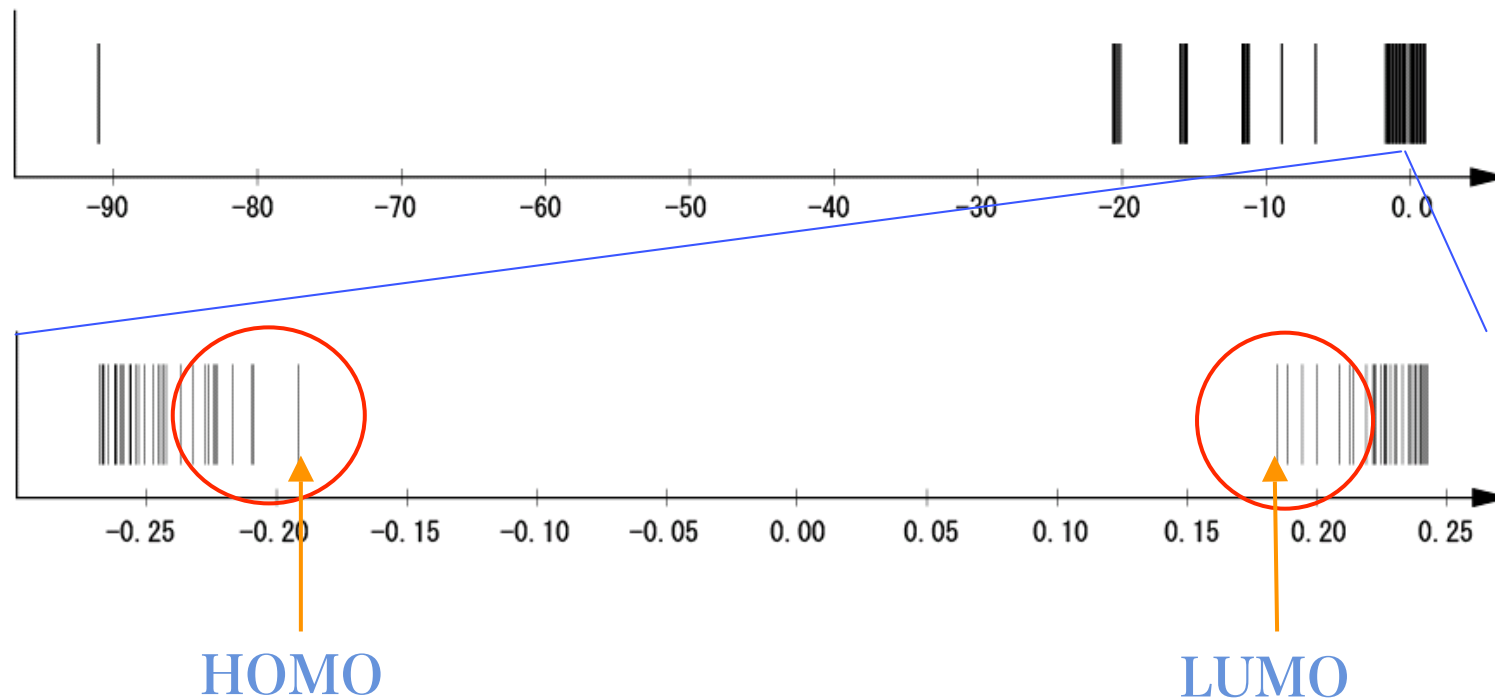
FMO-MO method



(large-scale eigenproblem)

Required Orbitals

Energy state:



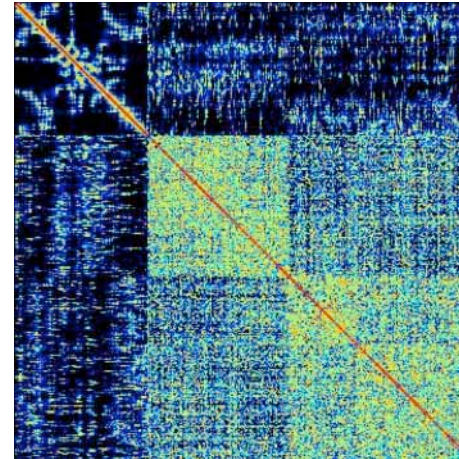
Eigenvectors related to chemical activities:

➡ Interior eigenvalue problems

Matrix Properties

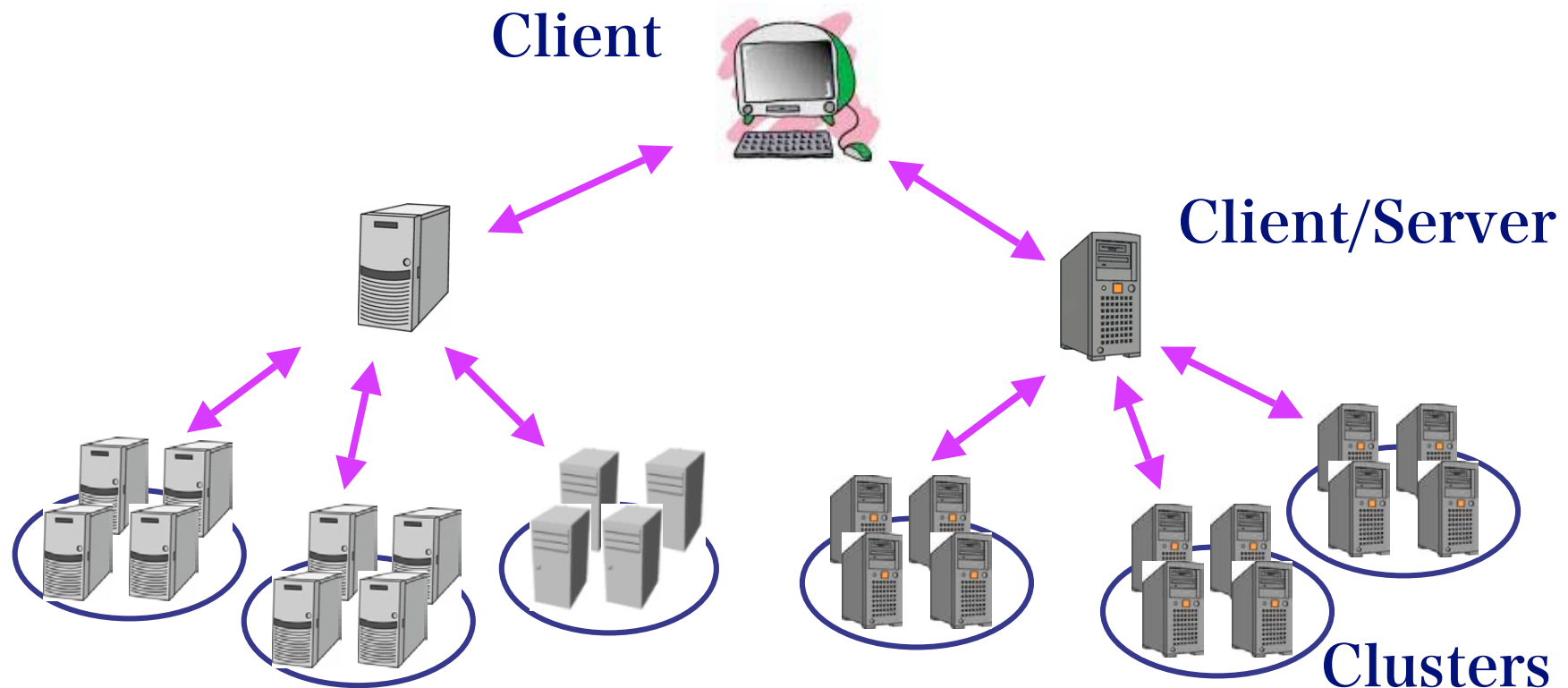
- The size of matrix: $2K \sim 200K$
The number of nonzero elements: $1M \sim 400M$
 - relatively large number of nonzero elements
 - unstructured sparsity pattern

Fock matrix of
Lysozyme + H₂O



Computing Environment

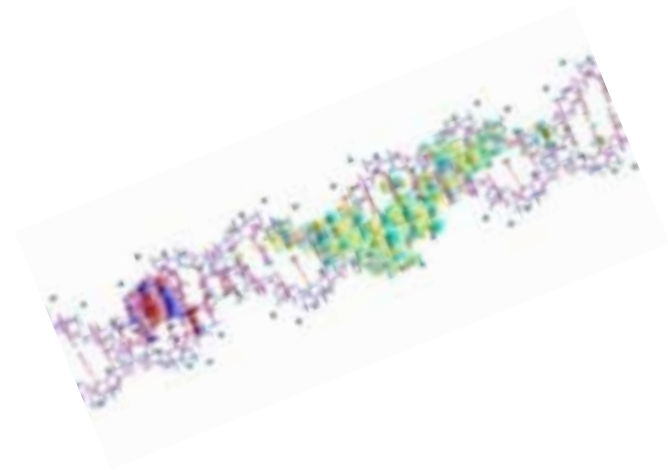
FMO-MO method is suitable for GRID computing.



Highly parallelized eigensolver is required.

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Generalized Eigenvalue Problem

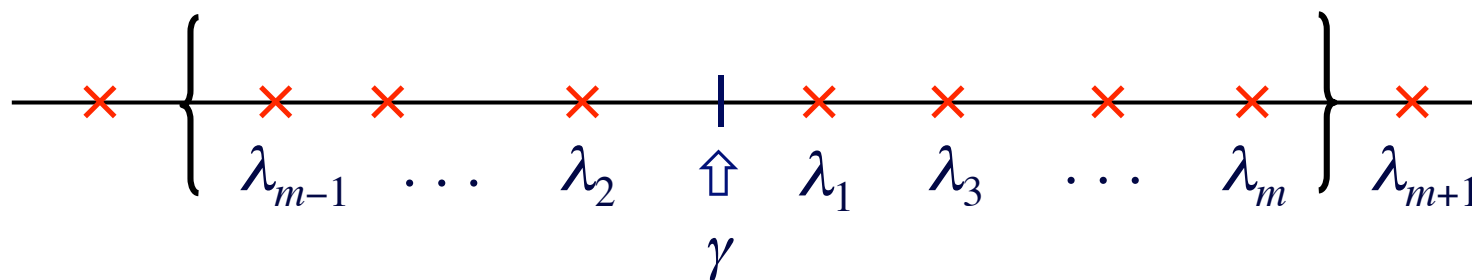
The generalized eigenvalue problem:

$$A\mathbf{x} = \lambda B\mathbf{x},$$

where $A, B \in \mathbb{R}^{n \times n}$, symmetric, and B is positive definite.

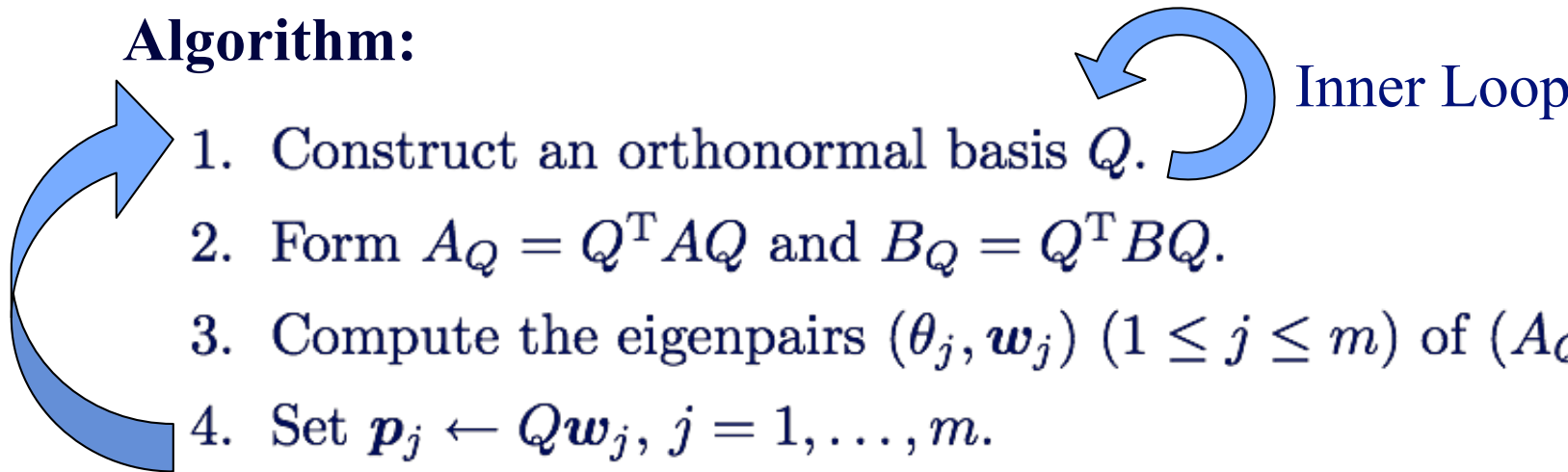
$(\lambda_j, \mathbf{u}_j)$: Eigenpair of the matrix pencil (A, B)

We find eigenpairs in a given interval:



Rayleigh-Ritz Procedure

Algorithm:

- 
1. Construct an orthonormal basis Q .
 2. Form $A_Q = Q^T A Q$ and $B_Q = Q^T B Q$.
 3. Compute the eigenpairs (θ_j, \mathbf{w}_j) ($1 \leq j \leq m$) of (A_Q, B_Q) .
 4. Set $\mathbf{p}_j \leftarrow Q \mathbf{w}_j$, $j = 1, \dots, m$.

Outer Loop

(A_Q, B_Q) : Projected pencil

θ_j : Ritz value

\mathbf{p}_j : Ritz vector

Contour Integral of Resolvent

To avoid inner/outer loops, we use a contour integral in construction of a subspace.

For a nonzero vector v , let

$$s_k := \frac{1}{2\pi i} \int_{\Gamma} (z - \gamma)^k (zB - A)^{-1} Bv dz,$$

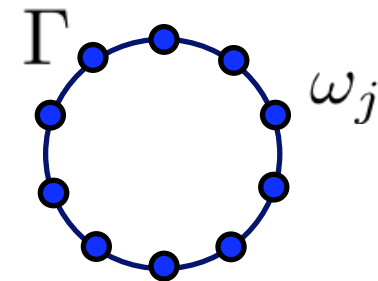
where $\Gamma \in \mathbb{C}$ is a Jordan curve that includes $\lambda_1, \dots, \lambda_m$.

$$\text{span}(s_0, \dots, s_{m-1}) = \text{span}(u_1, \dots, u_m)$$

[S and Tadano (2007)]

Approximation for Contour Integral

Γ : Circle with center γ and radius ρ



Equidistributed points on the circle:

$$\omega_j = \gamma + \rho e^{\frac{2\pi i}{N}(j+1/2)}, \quad j = 0, 1, \dots, N-1$$

s_k are approximated by the N -point trapezoidal rule:

$$\mathbf{s}_k \approx \frac{1}{N} \sum_{j=0}^{N-1} (\omega_j - \gamma)^{k+1} \mathbf{y}_j, \quad k = 0, \dots, m-1$$

where

$$\mathbf{y}_j = (\omega_j B - A)^{-1} \mathbf{v}, \quad j = 0, \dots, N-1$$

Contour Integral Rayleigh-Ritz Method

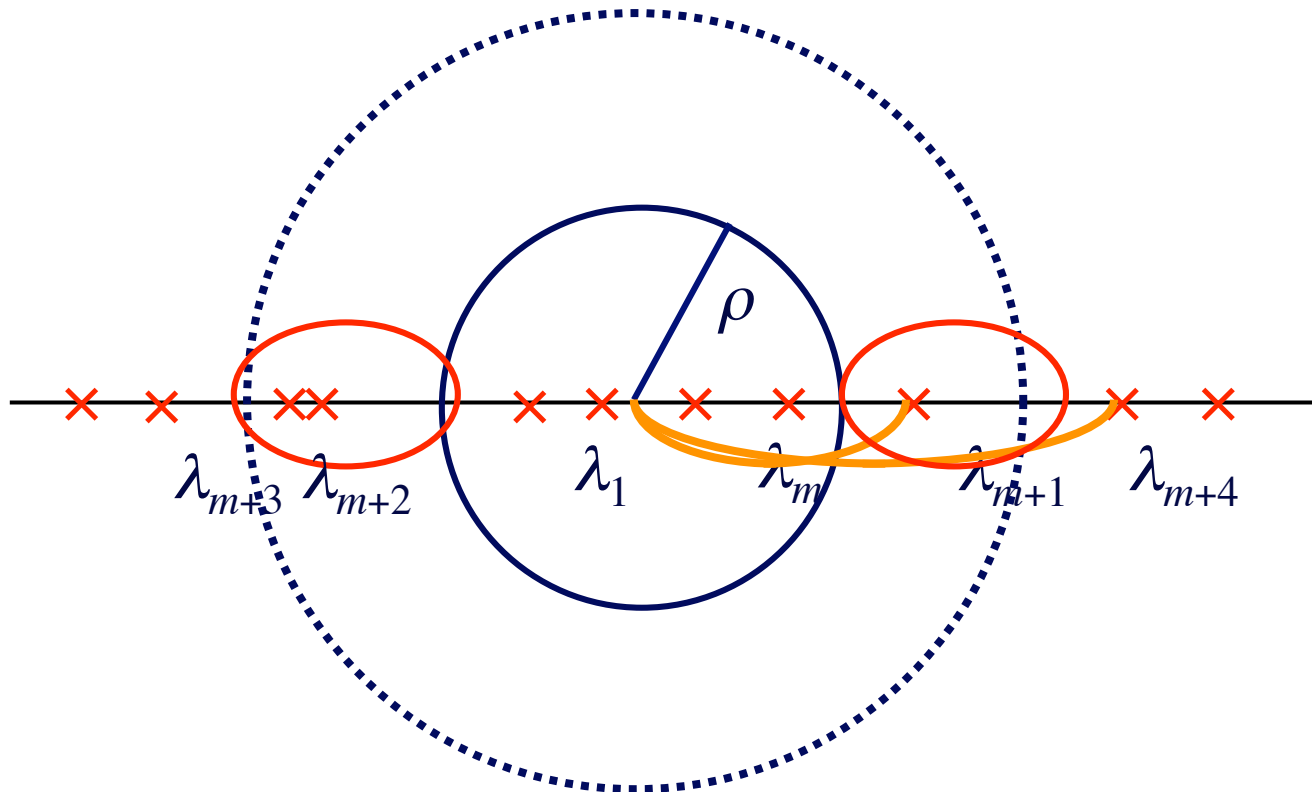
Algorithm of CIRR (Contour Integral Rayleigh-Ritz) method:

1. Set $\tau_j \leftarrow \exp(2\pi i(j + 1/2)/N)$, $j = 0, \dots, N - 1$.
2. Set $\omega_j \leftarrow \gamma + \rho \tau_j$, $j = 0, \dots, N - 1$. Construct a subspace
3. Solve $(\omega_j B - A)\mathbf{y}_j = \mathbf{v}$ for \mathbf{y}_j , $j = 0, \dots, N - 1$.
4. Set $\hat{\mathbf{s}}_k \leftarrow \sum_{j=0}^{N-1} \tau_j^{k+1} \mathbf{y}_j$, $k = 0, \dots, M-1$. Rayleigh-Ritz procedure
5. Construct an orthonormal basis Q from $\{\hat{\mathbf{s}}_0, \dots, \hat{\mathbf{s}}_{M-1}\}$.
6. Form $A_Q = Q^T(A - \gamma B)Q$ and $B_Q = Q^T B Q$.
7. Compute the eigenpairs (θ_j, \mathbf{x}_j) ($1 \leq j \leq M$) of (A_Q, B_Q) .
8. Set $\hat{\lambda}_j \leftarrow \theta_j + \gamma$, $j = 1, \dots, M$.
9. Set $\hat{\mathbf{u}}_j \leftarrow Q \mathbf{x}_j$, $j = 1, \dots, M$.

Influence of Quadrature Error

Let $\eta := \min_{j>m} \frac{\lambda_j - \gamma}{\rho} \quad \eta \rightarrow \min_{j>m'} \frac{\lambda_j - \gamma}{\rho}, \quad m' > m$

Then $|\hat{\lambda}_j - \lambda_j| = O(\eta^{-N}), \quad 1 \leq j \leq m$



Block Method

Block variant is also obtained by using a matrix

$$V = [\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_L]$$

instead of a vector \mathbf{v} .

$$\mathbf{s}_k := \frac{1}{2\pi i} \int_{\Gamma} (z - \gamma)^k (zB - A)^{-1} B \mathbf{v} dz$$

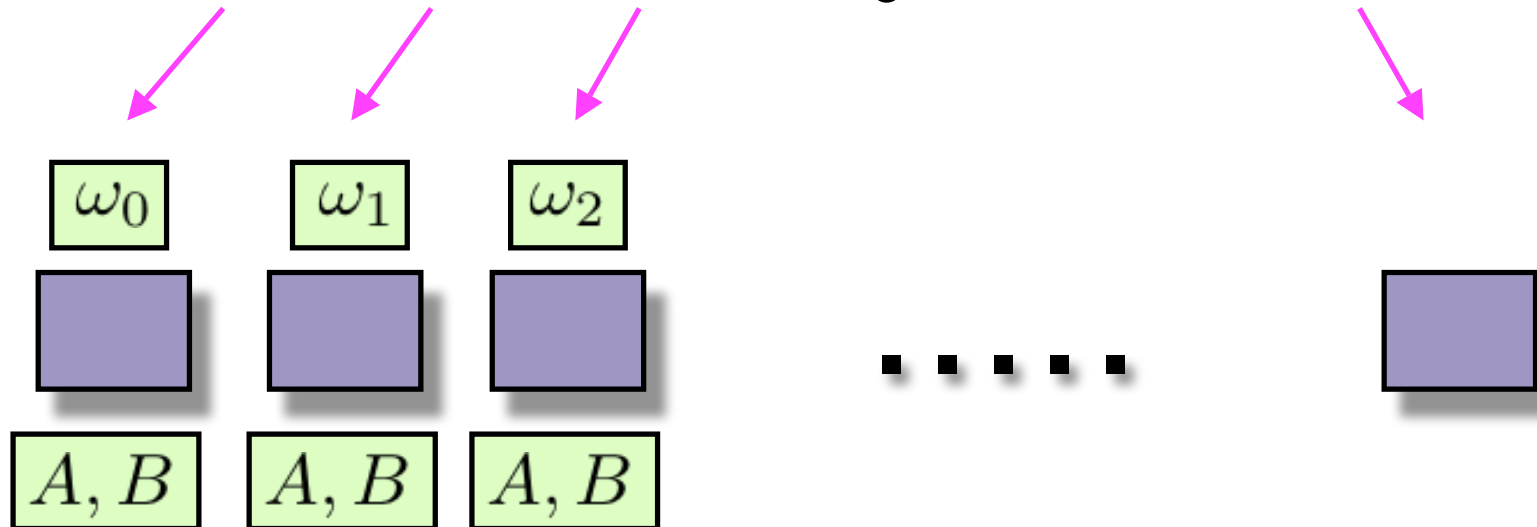
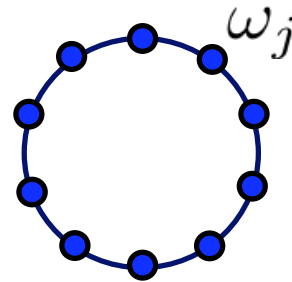


$$S_k := \frac{1}{2\pi i} \int_{\Gamma} (z - \gamma)^k (zB - A)^{-1} B V dz$$

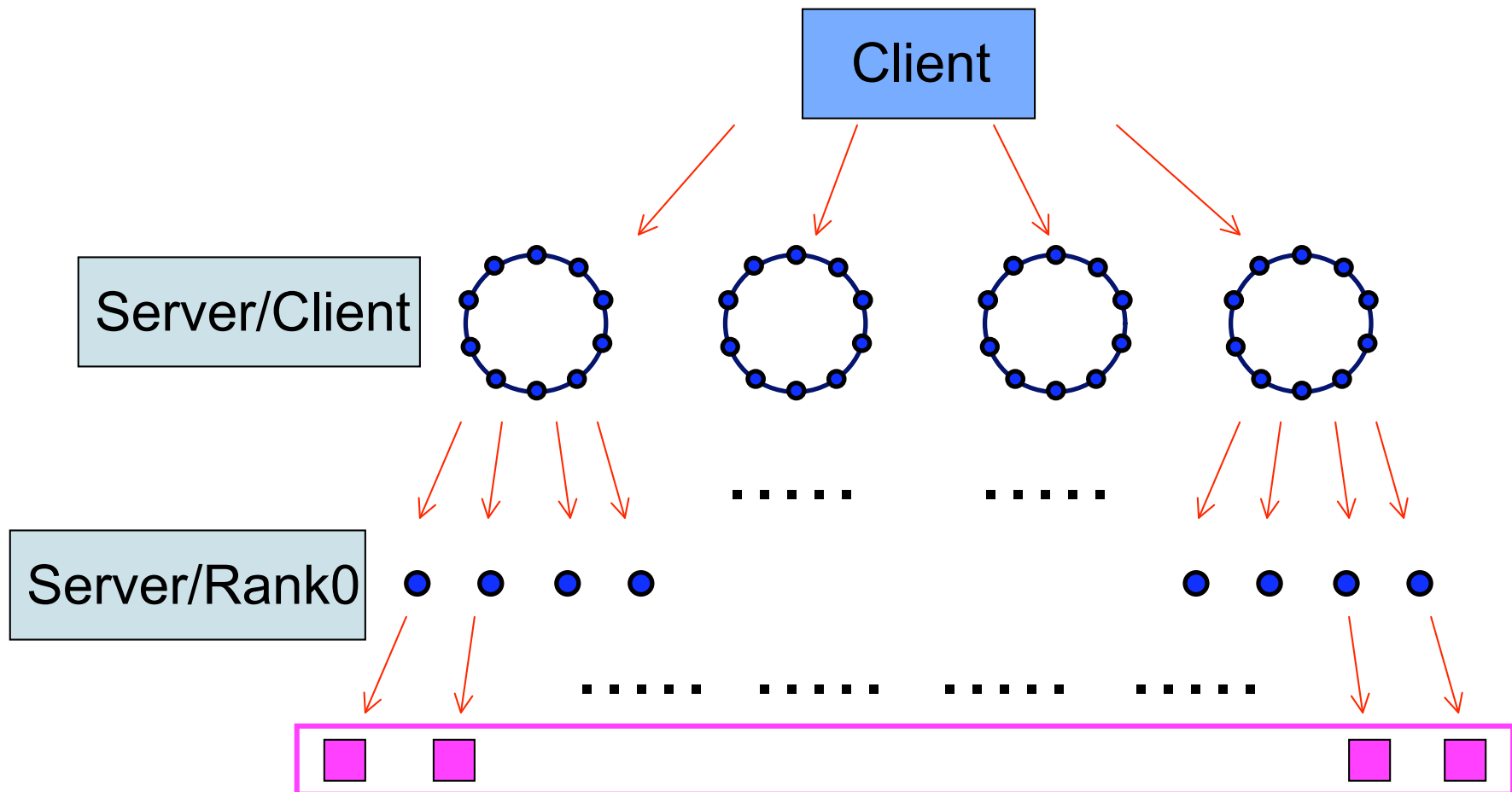
Parallel Implementation

$$\mathbf{s}_k \approx \frac{1}{N} \sum_{j=0}^{N-1} (\omega_j - \gamma)^{k+1} \mathbf{y}_j$$

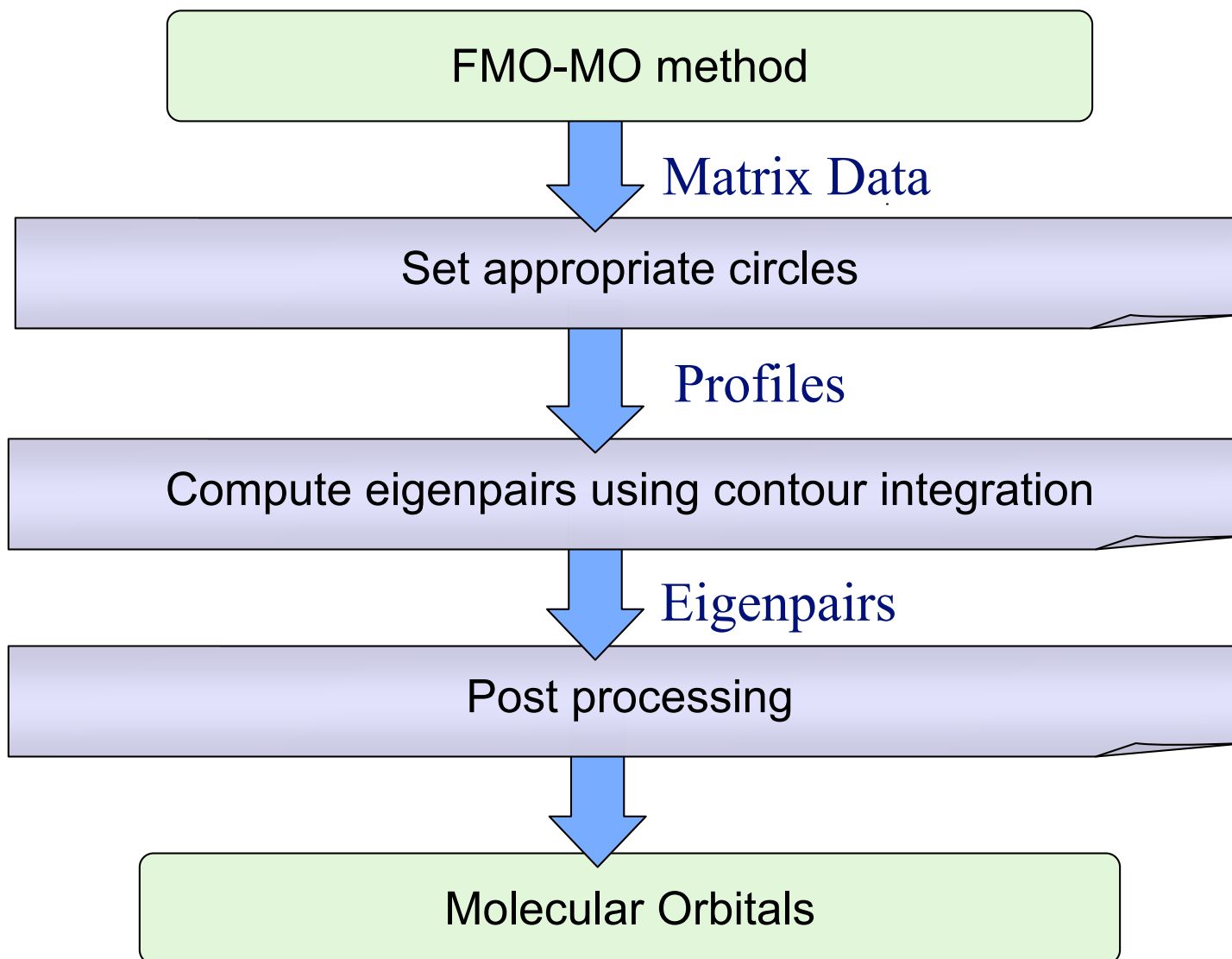
$$\mathbf{y}_j = (\omega_j B - A)^{-1} \mathbf{v}$$



Parallel Implementation

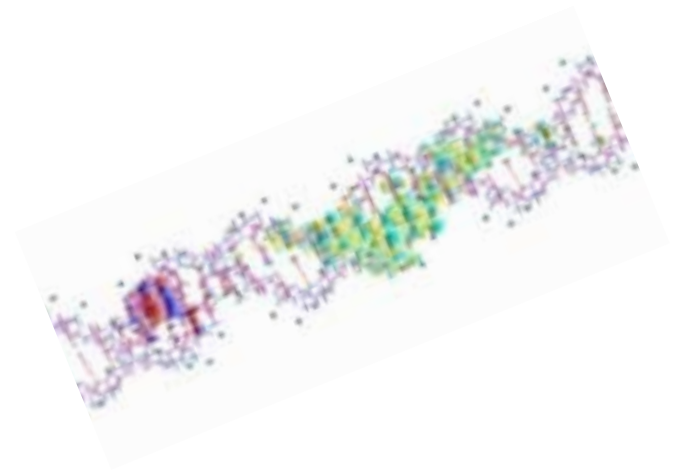


Flow of the Eigensolver



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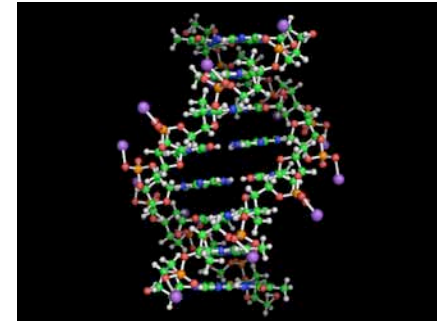
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Numerical Example (1)

- Test problem:

- Model of 8 DNA base pairs
- Matrix size: **1,980** × 1,980
- nnz: **728,080**



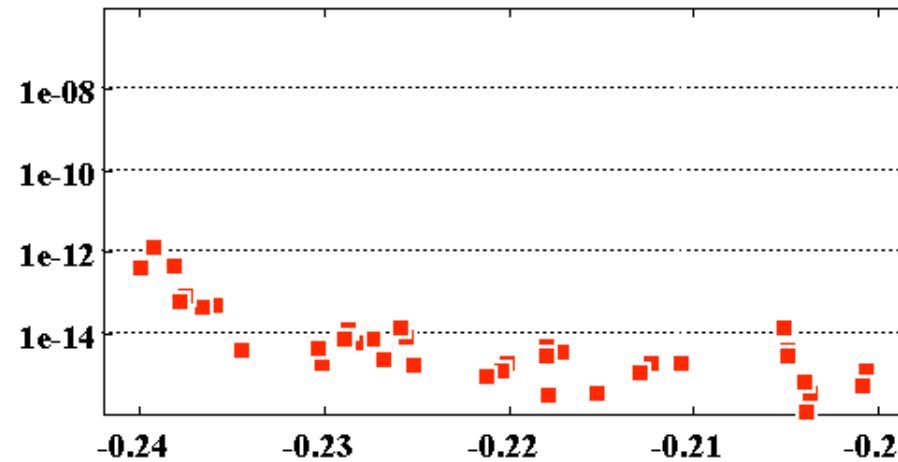
- Test Environment:

- OS: MacOSX 10.5
- CPU: **Core 2 Duo 2.2GHz** (2GB memory)
- Software: MATLAB 7.5
- Solver: UMFPACK (sparse direct solver)

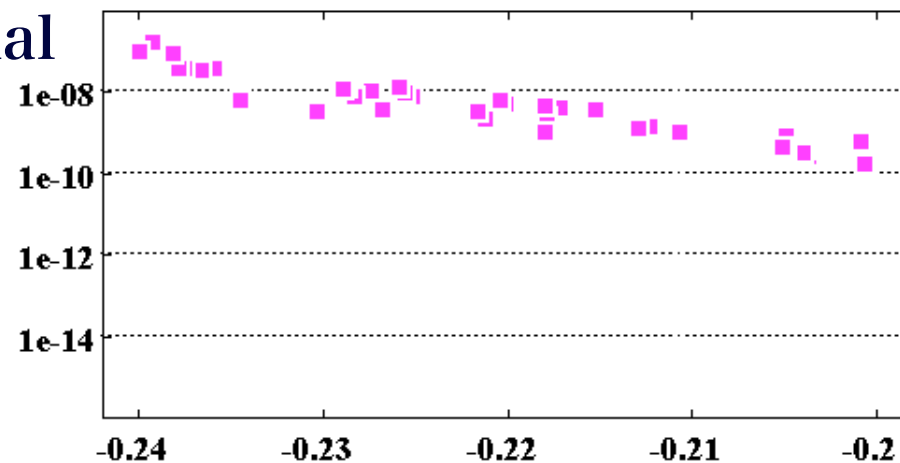
Numerical Example (1)

$L = 12$, $N = 16$, center = -0.22, radius = 0.02, 38 eigs

error

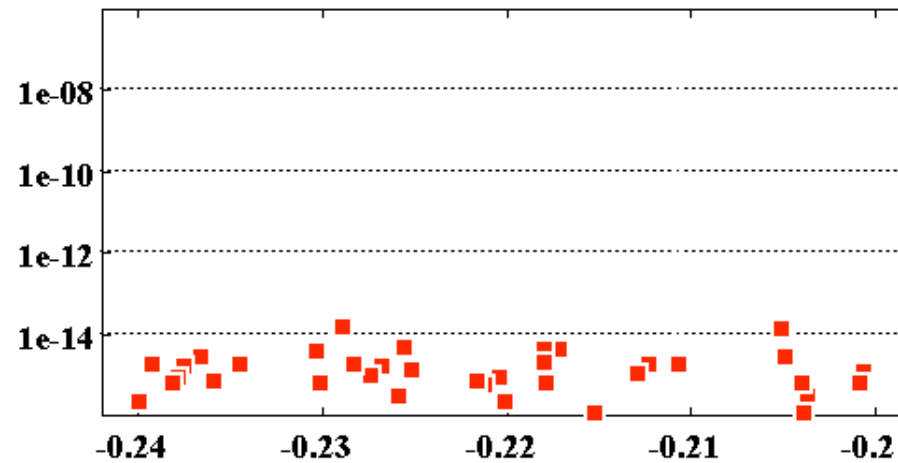


residual

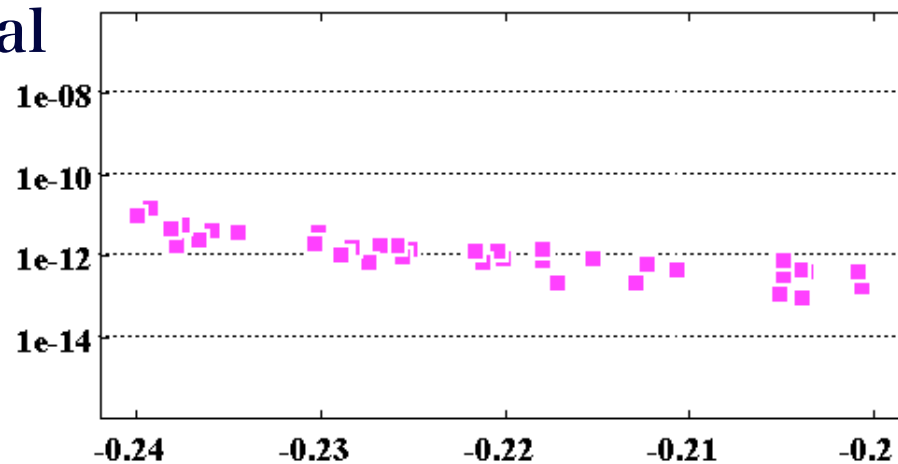


Numerical Example (1)

$L = 16$, $N = 24$, center = -0.22, radius = 0.02, 38 eigs
error

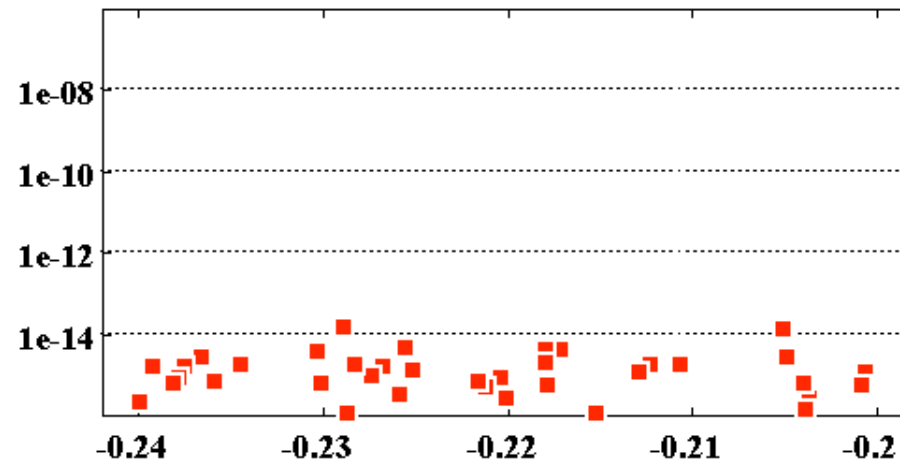


residual

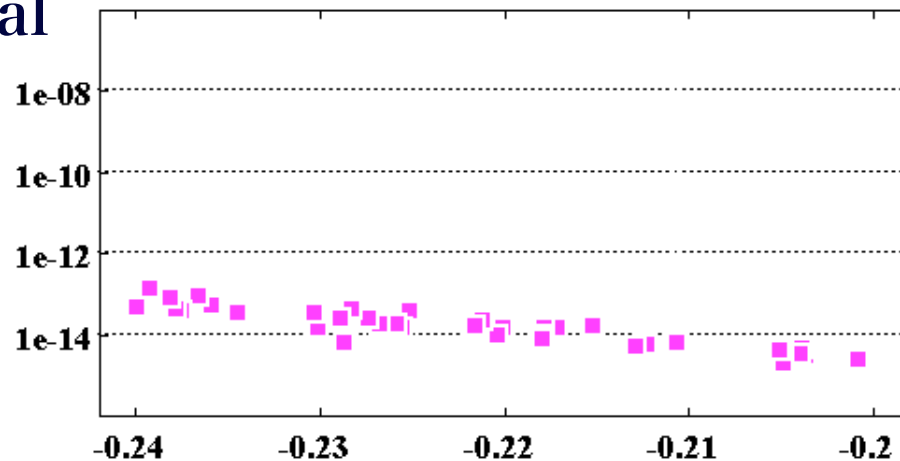


Numerical Example (1)

$L = 20$, $N = 24$, center = -0.22, radius = 0.02, 38 eigs
error



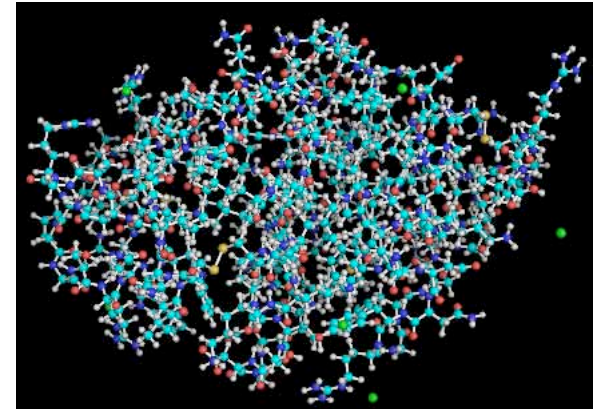
residual



Numerical Example (2)

- Test Problem:

- Lysozyme + H₂O
- Basis function: STO-3G
- Size: 20,758 × 20,758
- nnz: 20,064,444



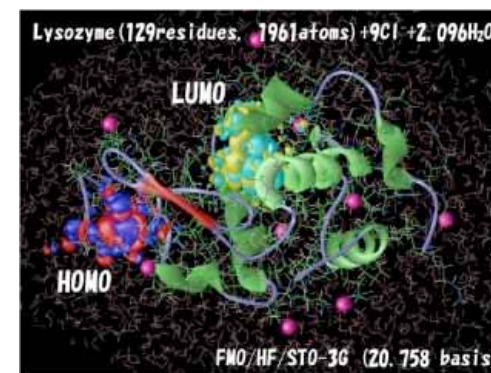
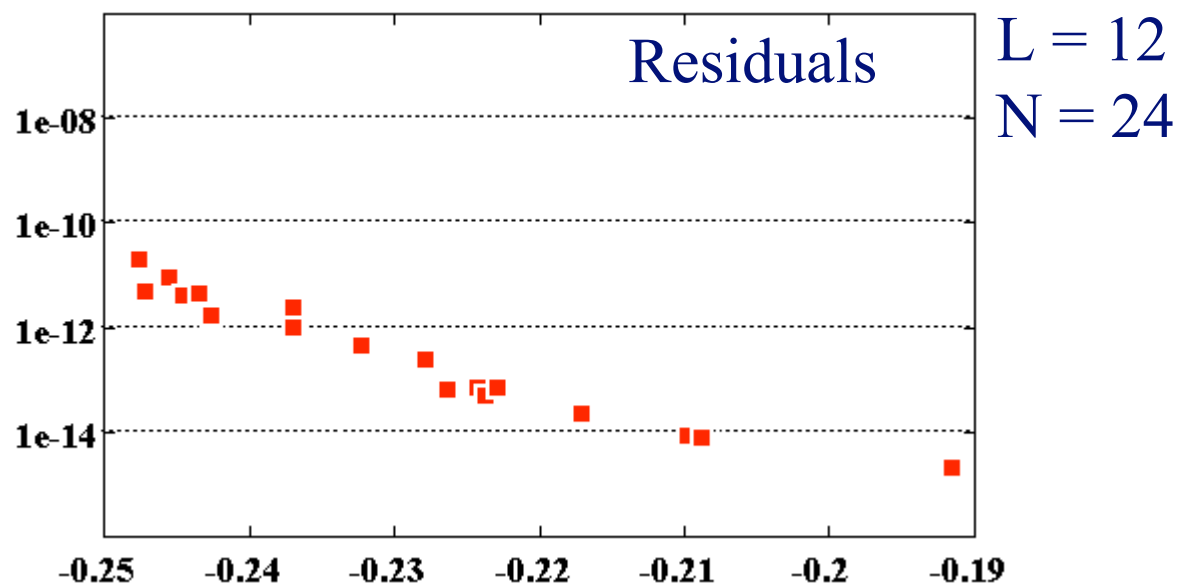
- Test Environment:

- OS: MacOSX 10.5
- CPU: Core 2 Duo 2.2GHz (2GB memory)
- Compiler: icc 10.1, ifort 10.1
- Solver: COCG method [van der Vorst and Melissen (1990)]
- Preconditioner: Complete Factorization for Approximate Matrix [Okada, S and Teranishi (2007)]
- Sparse Direct Solver for Preconditioner: PARDISO

Numerical Example (2)

Center: -0.22 Radius: 0.03 18 eigs

Wall-clock time: 233.2 sec

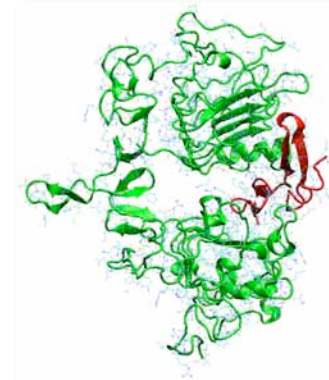


⌈ ARPACK+PARDISO: 316.1 sec, 20 eigs, max(res) = 6.6e-6 ⌋
(Xeon 3.2GHz 2MB Memory)

Numerical Example (3)

- Test Problems:

- EGF (Epidermal Growth Factor)
- Basis function: 6-31G
- Size: $43,612 \times 43,612$
- nnz: 73,175,935

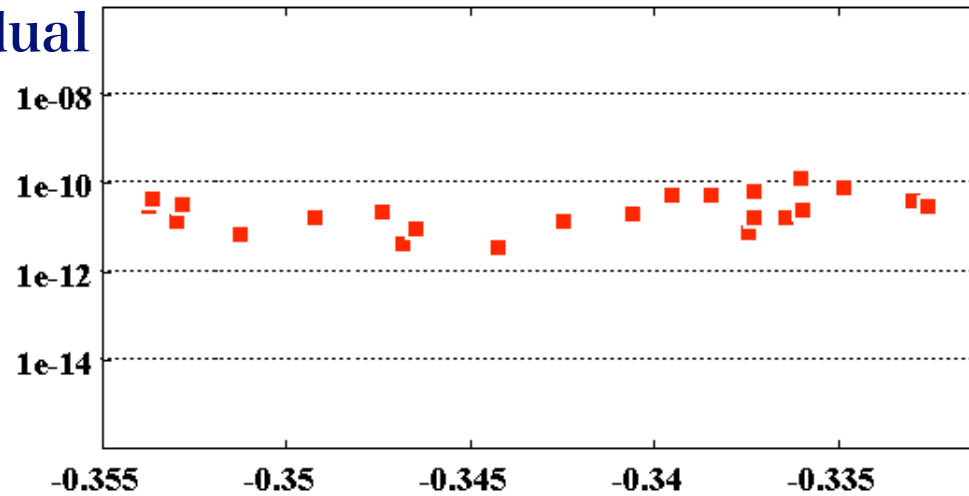


- Test Environment:

- OS: MacOSX 10.5
- CPU: Core 2 Duo 2.2GHz (2GB memory)
- Compiler: icc 10.1, ifort 10.1, MKL 10.0
- Solver: COCG method [van der Vorst and Melissen (1990)]
- Preconditioner: Complete Factorization for Approximate Matrix [Okada, S and Teranishi (2007)]
- Sparse Direct Solver for Preconditioner: PARDISO

Numerical Example (3)

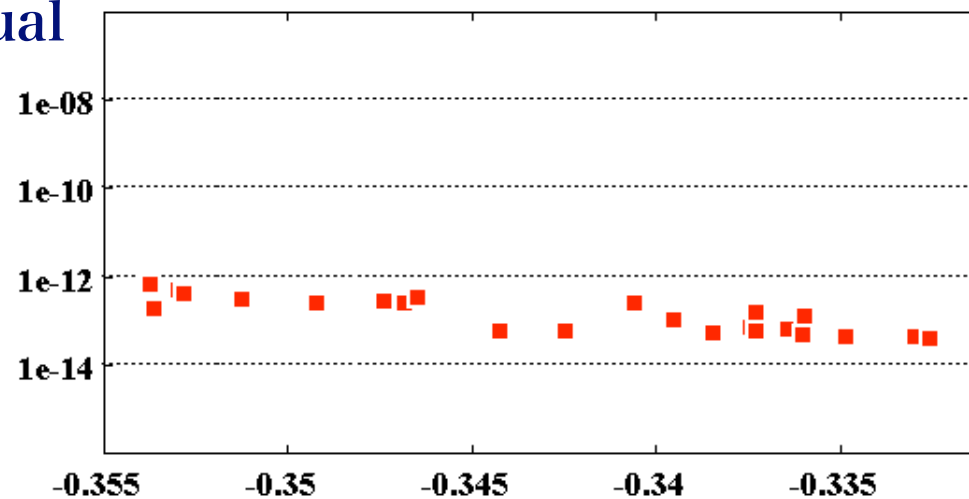
Residual



$L = 8$
 $N = 24$

1583.1 sec

Residual



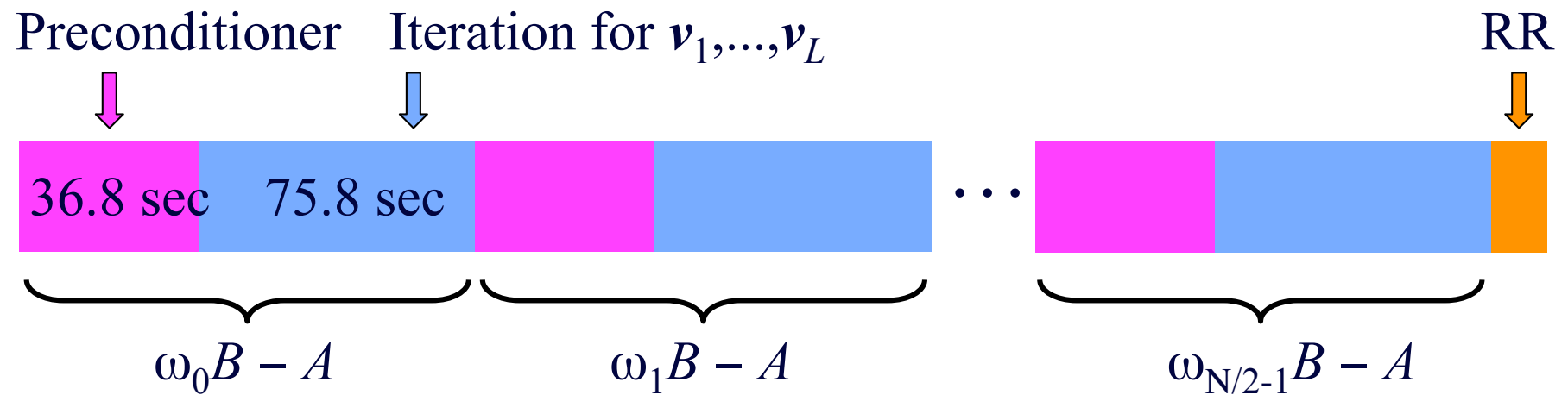
$L = 12$
 $N = 24$

2017.7 sec

Numerical Example (3)

Timing result (serial case):

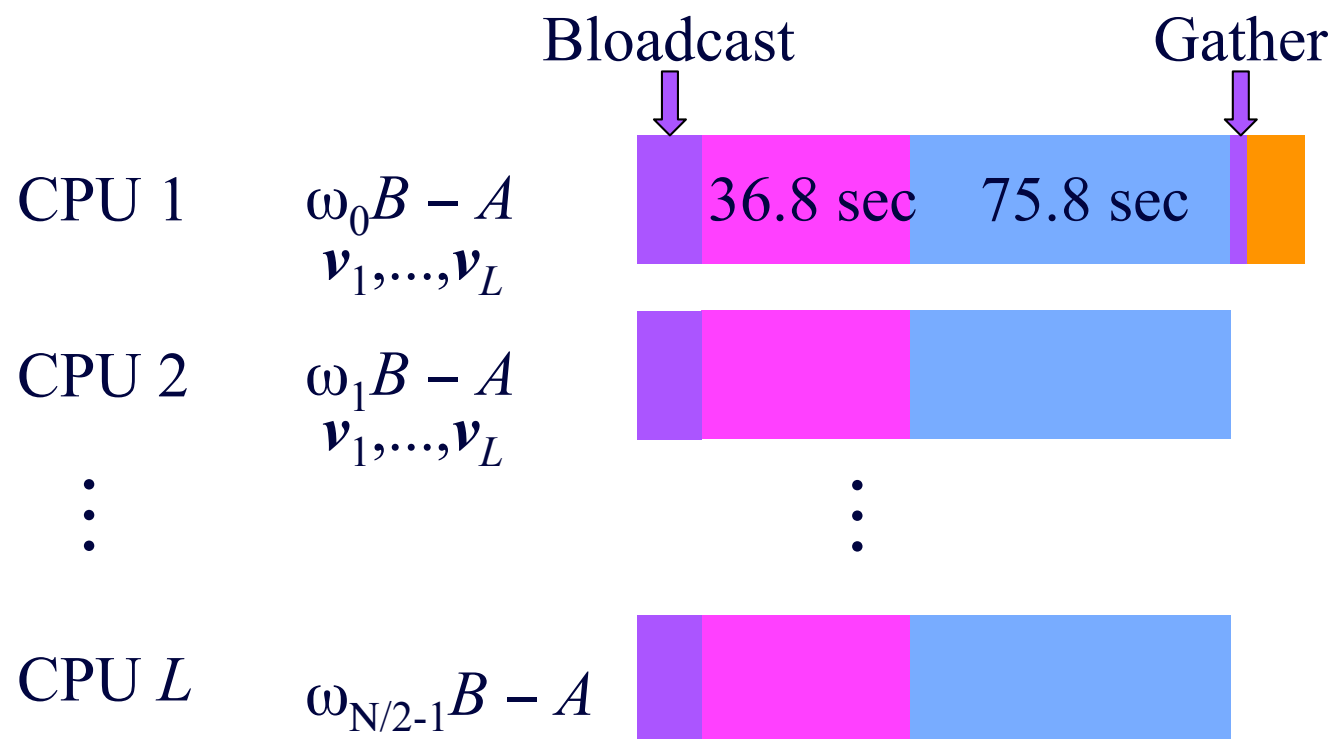
$L = 12$ $N = 24$ 2017.7 sec



Numerical Example (3)

Timing result estimation (parallel case 1):

$L = 12$ $N = 24$



Numerical Example (3)

Timing result estimation (parallel case 2):

$$L = 12 \quad N = 24$$

CPU 1

$$\omega_0 B - A$$

$$\mathbf{v}_1$$



CPU 2

$$\omega_0 B - A$$

$$\mathbf{v}_L$$



⋮

⋮

CPU $L^*(N/2)$

$$\omega_{N/2-1} B - A$$

$$\mathbf{v}_L$$



Summary

- A Rayleigh-Ritz type method using the contour integral was proposed.
- This method finds limited number of eigenpairs in a given interval.
 - Efficient for molecular orbital computation.
 - Easy to implement for distributed computing.
- Find good preconditioner.
- Application for other problems.
(Not only for SPD case)