

Improving robustness of the FEAST algorithm and solving eigenvalue problems from graphene nanoribbons

Lukas Krämer
GAMM Annual Meeting, Erlangen

Bergische Universität Wuppertal
—Applied Computer Science Group—

March 12th, 2014



Joint work with

- ▶ Martin Galgon
- ▶ Bruno Lang

from Bergische Universität Wuppertal,

- ▶ Andreas Alvermann
- ▶ Holger Fehske
- ▶ Andreas Pieper

from EMA Universität Greifswald.

This work was partially supported by DFG within the priority programme 1648—Software for Exascale Computing (SPPEXA) within the project ESSEX (Equipping Sparse Solvers for Exascale).

<http://www.sppexa.de>
<http://blogs.fau.de/essex/>



Outline

Introduction

Problem

The FEAST Algorithm

Graphene modeling

Numerical experiments

Graphene



Symmetric eigenvalue problem


We aim at solving the problem:

Given: $A \in \mathbb{R}^{n \times n}$, $A = A^T$,
an interval

$$I_\lambda = [\bar{\lambda}, \underline{\lambda}].$$

Sought: $X \in \mathbb{R}^{n \times m}$, $m \leq n$, $X^T X = I$,
 $\Lambda = \text{diag}(\lambda_1 \dots, \lambda_m)$ such that

$$AX = X\Lambda \text{ and all } \lambda_j \in [\bar{\lambda}, \underline{\lambda}].$$

-  Solve the (partial) symmetric eigenproblem. Find all eigenvalues in a given interval.

(Everything can be generalized to complex matrices and to a pair (A, B) with B hpd.)



... for solving eigenvalue problems (non-orthogonal version):

- ▶ Choose a subspace $\mathcal{U} = \text{span}(U)$
- ▶ Compute the Rayleigh quotients $A_U := U^H A U$, $B_U := U^H U$
- ▶ Compute the primitive Ritz pairs $(\tilde{\Lambda}, \tilde{W})$ of $A_U W = B_U W \Lambda$.
- ▶ Return the approximate Ritz pairs $(\tilde{\Lambda}, U \tilde{W})$ of $A X = X \Lambda$.

... and hope that $(\tilde{\Lambda}, U \tilde{W})$ are good approximations to some eigenpairs.



The subspace in use

[E. Polizzi, Phys. Rev. B, 2009, 79, 115112]

In FEAST: The subspace

$$U := \frac{1}{2\pi i} \int_{\mathcal{C}} (zI - A)^{-1} dz Y$$

is used.

- ▶ \mathcal{C} = curve in complex plane surrounding I_{λ} .
- ▶ Y = starting base of (possibly) random vectors.

It follows

$$\frac{1}{2\pi i} \int_{\mathcal{C}} (zI - A)^{-1} dz = \text{orth. projector onto } \text{span}(X)$$



The FEAST Algorithm: Skeleton

Input: An interval $I_\lambda := [\bar{\lambda}, \underline{\lambda}]$, an estimate \tilde{m} of the number of eigenvalues in I_λ .

Output $\hat{m} \leq \tilde{m}$ eigenpairs with eigenvalue in I_λ .

Perform:

1. Choose $Y \in \mathbb{C}^{n \times \tilde{m}}$ of rank \tilde{m} and compute

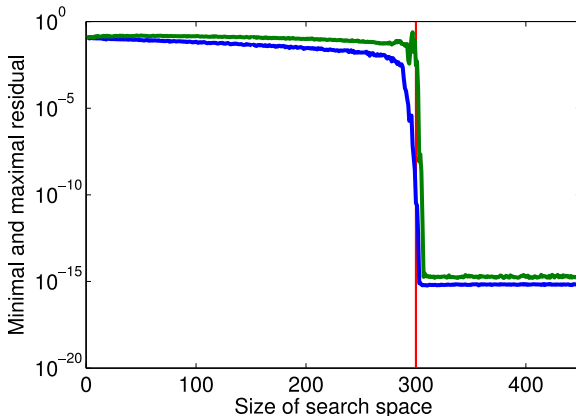
$$U := \frac{1}{2\pi\mathbf{i}} \int_{\mathcal{C}} (zI - A)^{-1} dz Y,$$

2. Form the Rayleigh quotients $A_U := U^H A U$, $B_U := U^H U$,
3. Solve the size- \tilde{m} generalized eigenproblem $A_U \tilde{W} = B_U \tilde{W} \tilde{\Lambda}$,
4. Compute the approximate Ritz pairs $(\tilde{\Lambda}, \tilde{X} := U \cdot \tilde{W})$,
5. If convergence is not reached go to Step 1, with $Y := \tilde{X}$.



Problem: How to choose \tilde{m} (initial dimension of \mathcal{U})?

Experiment: Choose $I_\lambda = [\bar{\lambda}, \underline{\lambda}]$ such that it contains the $m = 300$ lowest eigenvalues of size 1059-matrix. Let \tilde{m} vary, $\tilde{m} = 1, \dots, 450$
Residuals*:

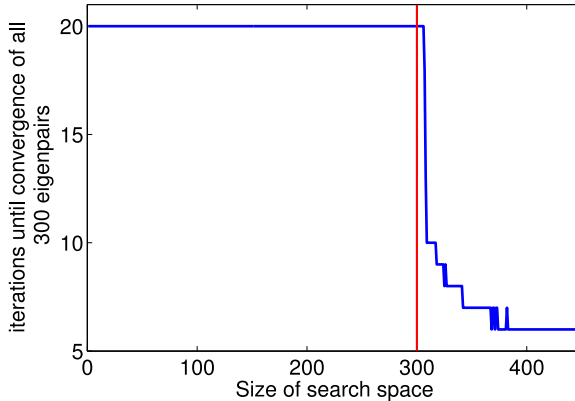


*K., Di Napoli, et al.: Dissecting the FEAST algorithm for generalized eigenproblems



Problem: How to choose \tilde{m} (initial dimension of \mathcal{U})?

Iteration counts[†]:



 Never choose \tilde{m} too small!

What if $\tilde{m} > m$ (results in too high numerical effort)?

[†]K., Di Napoli, et al.: Dissecting the FEAST algorithm for generalized eigenproblems



Counting eigenvalues

- ▶ $\text{rank}(U) = m =$ number of eigenvalues in I_λ
- ▶ In exact arithmetic: U has singular values 0 or 1
- ▶ numerical rank?



Counting eigenvalues

- ▶ $\text{rank}(U) = m =$ number of eigenvalues in I_λ
- ▶ In exact arithmetic: U has singular values 0 or 1
- ▶ numerical rank?
- ▶ It can be shown: If $m \leq \tilde{m}$ we have $m = \#$ of singular values of U that are $\geq 1/2$.
- ▶ Number $1/2$ depends on implementation (integration scheme). E.g., valid for Gauß–Legendre integration, midpoint rule of even order.



Counting eigenvalues: SVD of U

SVD $U = W\Sigma V^T$ with $\Sigma = \text{diag}(\sigma_1, \dots, \sigma_m, 0, \dots, 0)$

$\Rightarrow \text{span}(W(:, 1 : m)) = \text{span}(U)$

\Rightarrow Found the correct subspace, but cost = $\mathcal{O}(n^2 \cdot \tilde{m})$

SVD of $B_U = U^T U$:

$\Rightarrow B_U = V\Sigma^T \Sigma V^T$ with $V \in \mathbb{R}^{\tilde{m} \times \tilde{m}}$, orthogonal. **No order n cost.**

$\Rightarrow U' = U \cdot V(:, 1 : m)$ spans the wanted space $\rightsquigarrow \mathcal{O}(n \cdot m)$.



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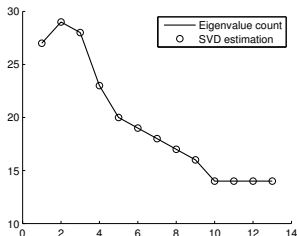
This means the dimension and a basis of the search space can be computed.

Proceed with a subspace slightly larger than m .

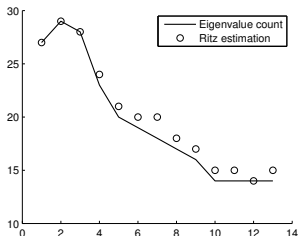


Counting eigenvalues: example

SVD count




Number of Ritz values in I_λ



1 Point on abscissa \equiv 1 interval I_λ

○ = count from respective method

line = exact count

 Ritz count is almost for free and not too bad. It does **not** deliver the subspace. Typically overestimates.



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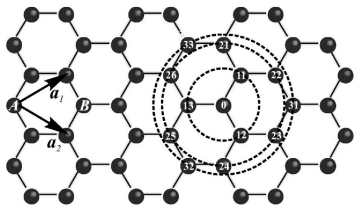
Numerical experiments

Graphene

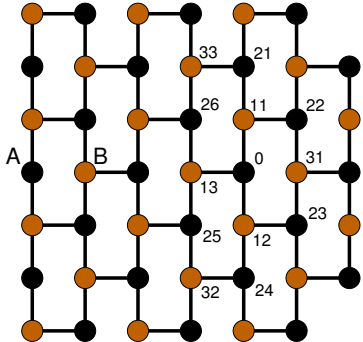


Graphene nanoribbons. . .

Hexagonal lattice

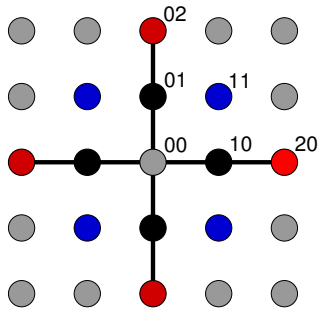
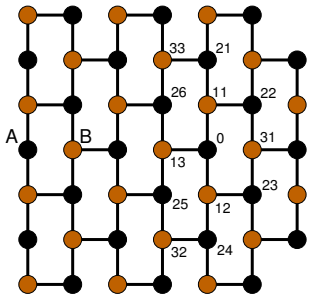


Brick lattice



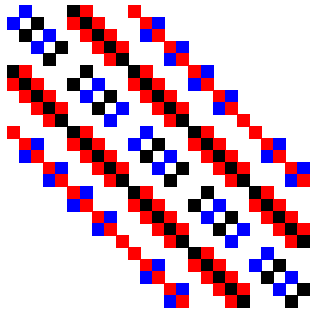
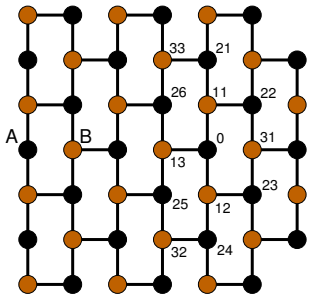
Graphene nanoribbons. . .

Model connections between neighbored atoms. . .



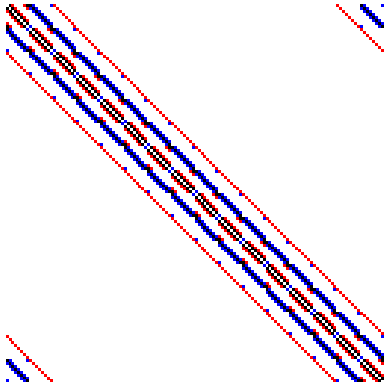
Graphene nanoribbons. . .

Model connections between neighbored atoms. . .

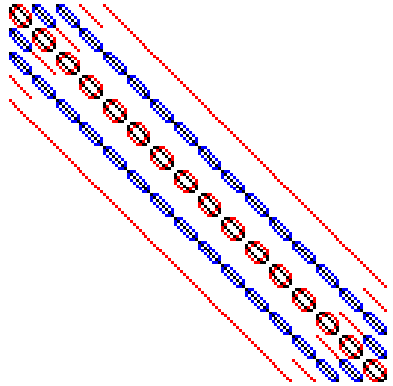


... lead to sparse eigenvalue problems

Graphene 8×16 , periodic BC



reordering

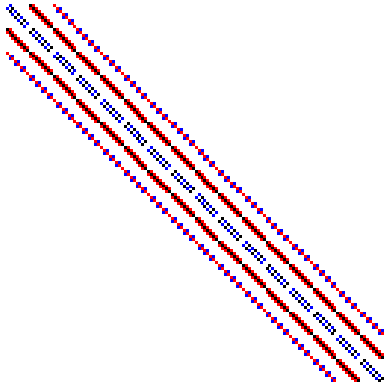


Nearest neighbor, next nearest neighbor, second nearest neighbor

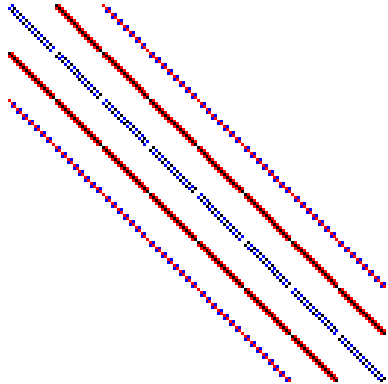


... lead to sparse eigenvalue problems

Graphene 8×16



Graphene 16×8

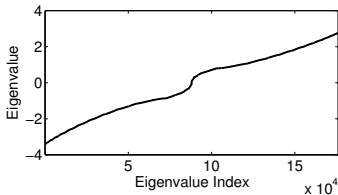


Nearest neighbor, **next nearest neighbor**, **second nearest neighbor**
 \Rightarrow sparse, symmetric matrix, width of graphene strip is reflected in bandwidth.



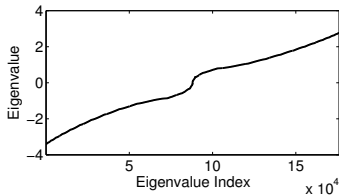
Wanted: Some eigenpairs with eigenvalue close to zero.

Graphene matrix of size 176,000:

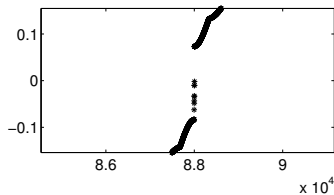


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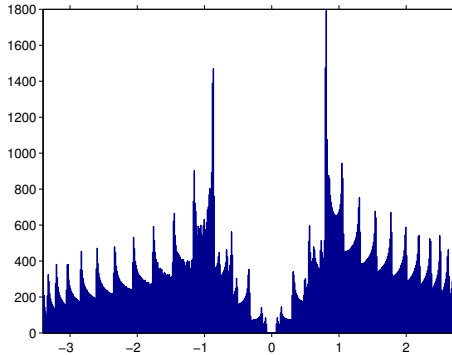
Graphene matrix of size 176,000:



Zoom on eigenvalues around zero:



Spectral density



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- ▶ Parallel C/MPI implementation incorporating Feast with the discussed eigenvalue counting

- ▶
$$U = \frac{1}{2\pi\mathbf{i}} \int_{\mathcal{C}} (zI - A)^{-1} dz Y \approx \sum_{j=0}^p \omega_j (z_j I - A)^{-1} Y$$

(Gauß–Legendre integration).

- ▶ 8 points used
- ▶ Solution of $(z_j I - A)V = Y$ via a partitioning method[‡] and recursive doubling (Hockney-Golub) for banded systems.

[‡]D. H. Lawrie, A. H. Sameh, The computation and communication complexity of a parallel banded system solver, ACM TOMS 10(2), 1984.



Results: Counting and residuals

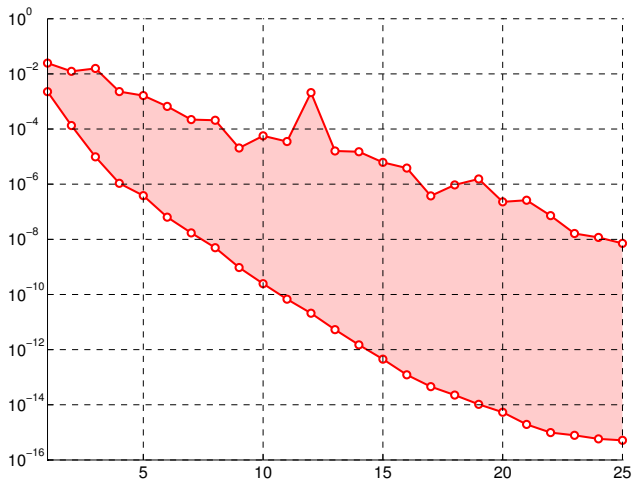
- ▶ Matrix size $n \approx 1.2M$, ratio $\frac{\text{nnz}}{n^2} \approx 0.001\%$ (12 nonzero entries per row).
- ▶ Graphene II with gap around zero.

Matrix	est.	found	iters.
Graphene I	535	535	7
▶ Graphene II (ℓ)	492	492	4
Graphene II (r)	546	546	8
Graphene III	574	572	25

- ▶ Eigenpair residuals: $10^{-9} \dots 10^{-16}$
- ▶ Linear system residuals: $10^{-10} \dots 10^{-16}$



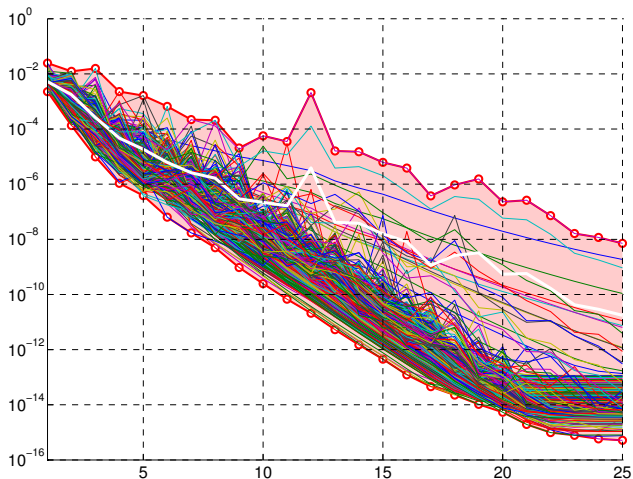
Residuals: history



Range of residuals vs. iteration number for Graphene III



Residuals: history



Individual residuals vs. iteration number for Graphene III.
White line = average.



Results: orthogonality

- ▶ measure $\text{orth} := \max_{i \neq j} |x_i^T x_j|$
- ▶ Problem: orthogonality across intervals
- ▶ Better if eigenvalues are separated
- ▶ For Graphene II (gap): $\text{orth} = 6 \cdot 10^{-11}$
- ▶ Individual intervals: $\text{orth} = 10^{-13} \dots 10^{-15}$ (basically ensured by design of algorithm).



Conclusion

- ▶ Reliable eigenvalue counting and
- ▶ efficient computation of a reasonable search space is possible



- ▶ Reliable eigenvalue counting and
- ▶ efficient computation of a reasonable search space is possible
- ▶ FEAST applied to eigenvalue problems from graphene yields accurate results.
- ▶ For certain problems, orthogonality issues are not very serious



Thank you for listening!

References:

- ▶ Krämer, L., Di Napoli, E., Galgon, M., Lang, B. and Bientinesi, P., Dissecting the FEAST algorithm for generalized eigenproblems, *J. Comput. Appl. Math.* **244**, pp. 1–9, 2013.
- ▶ Galgon, M., and Krämer, L., and Lang, B., Counting eigenvalues and improving the integration in the FEAST algorithm, Preprint BUW-IMACM 12/22
- ▶ Di Napoli, E., Polizzi, E., and Saad, Y., Efficient estimation of eigenvalue counts in an interval, Preprint 2013.
- ▶ Tang, P., and Polizzi, E., Subspace Iteration with Approximate Spectral Projection, Preprint 2013.

