

High-Performance Matrix Computations Final Projects

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Deadline: Midnight, the evening before your oral exam



● Routines

r1. Generalized eigensolvers	$\begin{cases} \text{xSYGV, xSYGVD, xSYGVX} & x \in [s d] \\ \text{xHEGV, xHEGVD, xHEGVX} & x \in [c z] \end{cases}$
r2. Dense eigensolvers	$\begin{cases} \text{xSYEVR, xSYEV, xSYEVD, xSYEVX} & x \in [s d] \\ \text{xHEEVR, xHEEV, xHEEVD, xHEEVX} & x \in [c z] \end{cases}$
r3. Tridiagonal eigensolvers	$\text{xSTEMR, xSTEQR, xSTEDC, xSTEVX} \quad x \in [s d]$

● Matrix size

n1. $n \in [10, \dots, 200]$

n2. $n \in [200, \dots, 1500]$

n3. $n \in [1500, \dots, N]$

$$N = \begin{cases} 10000 & \text{for r3} \\ 7000 & \text{for r2} \\ 4000 & \text{for r1} \end{cases}$$

● Matrix types

t1. Random entries, uniform distribution (0,1)

t2. Random entries, std normal distribution

t3. Random eigenvalues, uniform (0,1)

t4. Random eigenvalues, std normal distr.

t5. Uniform eigenvalue distribution*

t6. Geometric eigenvalue distribution*

t7. $n - 1$ eigvalues are ϵ , one is 1

t8. one eigvalue is ϵ , $n - 1$ are 1

t9. Matrix 1-2-1*

t10. Wilkinson-type matrix*

t11. Clement-type matrix*

t12. Legendre-type matrix*

t13. Laguerre-type matrix*

t14. Hermite-type matrix*

* : See page 119 of <http://arxiv.org/pdf/1401.4950v1.pdf>

Consider the eigensolvers in the group [r?] for datatype [?].

$$\text{r1: } Az = \lambda Bz \quad \text{r2: } Ay = \lambda y \quad \text{r3: } Tx = \lambda x$$

- Q1.** Compare the eigensolvers in terms of accuracy, performance and scalability, over a set¹ of matrices of type [t?] and different size [n?]. Document and report.
- Q2.** For each solver, give one or more indicative breakdowns of the computation time in terms of the calls directly within the routine. Repeat for 1 and ≥ 8 cores. Identify the bottleneck(s).
- QB. Bonus** (difficult): Assess the rate at which the flops of the tridiagonal solvers are performed. Compare with the TPP.
- Q3. Groups r1 and r2:** Optimize the blocksize NB of the reduction to tridiagonal form. Repeat with 1 and ≥ 8 cores.
- Q4. Group r1:** Relate the accuracy of each generalized solver to that of the corresponding dense and tridiagonal ones.
- Q5. Everybody:** You are given a sequence of at least 50 random symmetric tridiagonal matrices (type t1) of fixed size². The solver of choice is xSTEDC. What is the best way to make use of the available cores? Motivate your reasoning.

¹At least 6 matrices per matrix type.

²Fix a size $\bar{n} \in [n?]$.

How to build a matrix M with a given eigenspectrum?³

- The idea is to use similarity transformations, as they preserve the eigenspectrum. If Q is a dense orthogonal matrix, then $A := Q * \Lambda * Q^H$ is dense and $\lambda(A) = \lambda(\Lambda)$.
- For a dense eigenproblem, first construct Λ and then compute A .
In order to create a dense orthogonal matrix Q , compute the QR factorization of a random matrix M .
- For a generalized eigenproblem, first build a dense problem, then construct an SPD matrix B , compute its Cholesky factor L , and $A := L * A * L^H$.

Accuracy for a generalized eigenproblem

- The eigenvectors X of the generalized eigenproblem $Ax = \lambda Bx$ are B -orthogonal: $X^H * B * X = I$.

³See attached file.

- Execute on one of the compute nodes of RZ's cluster:
<https://doc.itc.rwth-aachen.de/display/CC/Hardware+of+the+RWTH+Compute+Cluster>
- Possible approaches are manual instrumentation, ELAPS, profilers and tools such as gprof and VTune, and any combination of them.
- Submit all your work, code, scripts, makefiles.
- For Q1, submit the tridiagonal representation of the matrices used.
- Prepare a report (pdf,org,html). Describe mathematically your computations.
- Archive the files: `your_name.tgz` or `your_name.zip`
- Submission by email to `pauldj@aices.rwth-aachen.de`
- Email's subject: 'HPMC-15 Project your_last_name'
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