

# 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  $\epsilon$ , one is 1

t8. one eigvalue is  $\epsilon$ ,  $n - 1$  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 set<sup>1</sup> 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  $\geq 8$  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  $\geq 8$  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 size<sup>2</sup>. The solver of choice is xSTEDC. What is the best way to make use of the available cores? Motivate your reasoning.

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<sup>1</sup>At least 6 matrices per matrix type.

<sup>2</sup>Fix a size  $\bar{n} \in [n?]$ .

## How to build a matrix $M$ with a given eigenspectrum?<sup>3</sup>

- 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  $\Lambda$  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:  $X^H * B * X = I$ .

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<sup>3</sup>See 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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