

# ExaNLA Survey Response Report

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## Submission Details

Library Name: SIESTA  
Version: 5.4.1  
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Organization: ICMAB-CSIC

## Selected NLA Operations

1. Cholesky Factorization

### 1. Codes Information

*Basic information about your application/simulation codes.*

Library Name:  
**SIESTA**

Current Version:  
**5.4.1**

Contact Information:  
**Not specified**

Name:  
**Alberto Garcia**

Email:  
**albertog@icmab.es**

Organization:  
**ICMAB-CSIC**

Application Domain:  
**Not specified**

What is the primary application domain of your codes?:  
**Materials Science**

Materials Science:  
**Not specified**

What are the main functionalities of your code?:  
**Ground state DFT, Time-dependent DFT, Molecular dynamics, Quantum transport, Crystal structure prediction, Phase transitions, Defect calculations, Surface science**

If you selected "Other", please specify::  
**Electrochemistry**

Climate/Weather Modeling:  
**Not specified**

What are the main functionalities of your code?:  
**Not specified**

If you selected "Other", please specify:  
**Not specified**

Fluid Dynamics:  
**Not specified**

What are the main functionalities of your code?:  
**Not specified**

If you selected "Other", please specify:  
**Not specified**

Other Domain Functions:  
**Not specified**

What are the main functionalities of your code?:  
**Not specified**

Use Case Information:  
**Not specified**

Does your codes have multiple distinct use cases?:  
**Yes, multiple distinct use cases**

Which use case are you describing in this submission?:  
**Ground state DFT calculations**

Library Description:  
**SIESTA is a DFT code using pseudopotentials and a basis set of strictly localized pseudo-atomic orbitals. It also has a sub-system that uses Non-Equilibrium Green's Functions to deal with systems under bias (e.g. for electron transport and electrochemistry)**

## 2. Cholesky Factorization

Cholesky Factorization ( $A = LL^T$ ):  
**Yes**

Diagonal Dominance:  
**Weakly diagonally dominant**

Condition Number:  
**Moderately conditioned ( $10^3 - 10^6$ )**

Matrix Properties and Structure:  
**Dense**

Matrix Distribution:  
**Block cyclic distribution (e.g., ScaLAPACK style)**

Matrix Storage Format:  
**Dense (column-major/row-major)**

Matrix Dimensions:  
**Large (10,000 – 100,000)**

Factorization Tolerance:  
**Very high accuracy ( $10^{-12}$ )**

Working Precision:  
**Double precision (64-bit)**

Workload Characteristics:  
**Not specified**

Computation Pattern: capability or capacity:  
**Large-scale single factorizations (e.g., one large matrix at a time, using significant computational resources), Many independent smaller factorizations (e.g., batch processing multiple matrices simultaneously), Repeated factorizations of similar matrices (e.g., during iterative refinement or optimization), Part of larger computation (e.g., reduction of generalized eigenproblems)**

Distributed-Memory Dense NLA Library Usage:

**Not specified**

Currently Used Libraries:

**ScaLAPACK, ELPA**

Interested in Using, but not currently using:

**SLATE, EigenExa, DLA-Future**

Specialized Libraries (Sparse/Structured/Hierarchical):

**Not specified**

Currently Used Libraries:

**Not specified**

Interested in Using, but not currently using:

**Not specified**

Benchmarking Requirements:

**Not specified**

Benchmark Input Types:

**Real matrices from application workloads, Mini-apps or extracted kernels from real applications**

Can You Provide Data or Mini-apps?:

**Yes, both matrices and mini-apps**

Scaling Requirements:

**Both strong and weak scaling needed**

### **3. Generalized Eigenvalue Problems ( $Ax = \lambda Bx$ )**

**Symmetric/Hermitian A, SPD B**

Matrix Structure:

**A is dense, B is dense, A is sparse, B is sparse**

Reduction to Standard Eigenproblem (using B):

**Not specified**

Reduction to Standard Eigenproblem:

**Yes, always**

Reduction Method:

**Cholesky factorization of B ( $B = LL^T$  or  $B = L^*L$ )**

Matrix Properties:

**Not specified**

Eigenvalue distribution:

**Mix of clustered and separated**

Problem Scale:

**Large (10,000 - 100,000)**

Computation Requirements:

**Not specified**

Percentage of eigenvalues:

**10-50%**

What to compute:

**Eigenvalues and eigenvectors**

Eigenvalue location:

**Smallest eigenvalues**

Required tolerance/precision:

**Not specified**

Residual tolerance type:  
**Other: I am not sure. This is handled by external libraries**

Absolute residual tolerance:  
**Not specified**

Relative residual tolerance:  
**Not specified**

Hybrid residual tolerance:  
**Not specified**

Orthogonality tolerance:  
**Very high ( $10^{-12}$ )**

Working Precision:  
**Single precision (32-bit), Double precision (64-bit)**

Workload Characteristics:  
**Not specified**

Computation Pattern: capability or capacity:  
**Large-scale single problems (e.g., one large generalized eigenproblem at a time, using significant computational resources), Many independent smaller problems (e.g., batch processing multiple generalized eigenproblems simultaneously), Repeated similar-sized problems (e.g., time evolution or parameter sweeps)**

Distributed-Memory NLA Library Usage:  
**Not specified**

Distributed-Memory Dense Linear Algebra:  
**ScaLAPACK, ELPA**

Iterative Eigensolvers:  
**Not specified**

High-Level & Interface Libraries:  
**ELSI – Abstraction layer for eigenvalue solvers (e.g., used by SIESTA, FHI-aims)**

Are there any NLA libraries you are interested in using (but have not yet adopted)?:  
**DLA-Future, ChASE, EigenExa, SLATE**

Benchmarking Requirements:  
**Not specified**

Benchmark Input Types:  
**Real matrices from application workloads, Mini-apps or extracted kernels from real applications**

Can You Provide Data or Mini-apps?:  
**Yes, both matrices and mini-apps**

Scaling Requirements:  
**Both strong and weak scaling needed**