









Electronic-Structure Solvers in SIESTA: Features and Performance Alberto García (ICMAB-CSIC, Barcelona)

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The basic core of SIESTA

$$\psi_{i}(r) = \sum_{\mu} \phi_{\mu}(r) c_{\mu i},$$

$$\sum_{\nu \beta} (H^{\alpha \beta}_{\mu \nu} - E_{i} S_{\mu \nu} \delta^{\alpha \beta}) c^{\beta}_{\nu i} = 0$$

 $ho(m{r}) = \sum_{\mu
u}
ho_{\mu
u} \phi^*_
u(m{r}) \phi_\mu(m{r})$

$$\rho_{\mu\nu} = \sum_{i} c_{\mu i} n_i c_{i\nu}$$

Density matrix

The SOLVER step takes most of the CPU time

Generalized eigenvalue problem

$$S_{\mu\nu} = \langle \phi_{\mu} | \phi_{\nu} \rangle$$





Need to use DIRECT solvers, as the number of desired eigenvectors is a substantial fraction of the matrix size

SIESTA uses pre-packaged libraries for this pure math problem:

- ScaLaPACK
 - pdsyev, pzheev and related drivers
 - MRRR
- ELPA: Alternative transformation sequence + optimizations https://elpa.mpcdf.mpg.de/

 $\sum_{\nu\beta} (H^{\alpha\beta}_{\mu\nu} - E_i S_{\mu\nu} \delta^{\alpha\beta}) c^{\beta}_{\nu i} = 0$

- Conversion of H and S to dense form
- Cholesky decomposition to reduce to standard eigenproblem
- Transformation to tri-diagonal form
- Solution of tri-diagonal problem
- Back-transformation

Cubic scaling with matrix size — Quadratic scaling for memory

Still competitive for low-cardinality basis sets



Direct solution for the density matrix

$$\hat{\rho} = f_{\beta}(\hat{H} - \mu)$$
$$f_{\beta}(\epsilon_i - \mu) = \frac{2}{1 + e^{\beta(\epsilon_i - \mu)}}$$

Fermi-Dirac function



Fermi Operator Expansion (FOE)

$$p(H) = \frac{c_0}{2}I + \sum_{j=1}^{n_{pl}} c_j T_j(H)$$

Calculation of the DM involves only (sparse) matrix-vector multiplications

CheSS library

(originally in BigDFT)

Linear-scaling



- Number of terms in the expansion can be large
- Efficiency increases for contracted basis sets.
- Exploring on-the-fly contraction



PEXSI: Pole Expansion plus Selected Inversion (Lin Lin, Chao Yang, et al., Berkeley)



$$\hat{\rho} = Im\left(\sum_{l=1}^{P} \frac{\omega_l}{H - (z_l + \mu)S}\right)$$



(Due to sparsity of the target density matrix)

Relatively small number of poles (20-30) Trivially parallelizable over them





ELSI initiative to integrate solver libraries



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GPU acceleration for diagonalization



ELSI-ELPA GPU acceleration

Future enhancements in ELPA (better kernels) and in ELSI (e.g. build-DM stage) are integrated in SIESTA automatically

Marconi-100 (CINECA): 32 CPUs+ 4 GPUs /node



Proper binding of GPUs to MPI ranks



System: Si quantum dot, with approx 35000 orbs



Massive scalability: PEXSI solver

$$\hat{\rho} = Im\left(\sum_{l=1}^{P} \frac{\omega_l}{H - (z_l + \mu)S}\right)$$

PEXSI offers:

- Three levels of parallelization (over orbitals, poles, and chemical potential values)
- A reduced memory footprint (only sparse matrices are stored)
- Reduced complexity (maximum O(N²) size scaling)





Deliverable D4.3 Second report on code profiling and bottleneck

Comparison of global efficiency of solvers for a very large problem

SARS CoV Approx 8

SARS CoV-2 M^{pro} with solvation water molecules

Approx 8800 atoms; 58000 orbitals





Work on GPU acceleration of PEXSI library is under way



THANKS





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