















Alberto García (ICMAB-CSIC, Barcelona)



The basic core of SIESTA

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

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

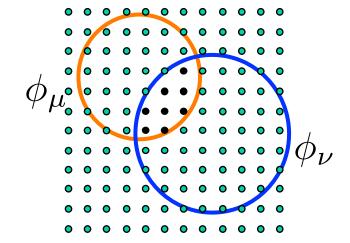
Generalized eigenvalue problem

$$\rho(\boldsymbol{r}) = \sum_{\mu\nu} \rho_{\mu\nu} \phi_{\nu}^*(\boldsymbol{r}) \phi_{\mu}(\boldsymbol{r})$$

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

Density matrix

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



The SOLVER step takes most of the CPU time



Diagonalization-based solvers

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

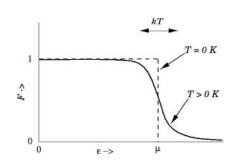


Direct solution for the density matrix

$$\hat{\rho} = f_{\beta}(\hat{H} - \mu)$$

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

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



Stephan Mohr (BSC)

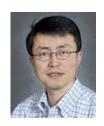
- Number of terms in the expansion can be large
- Efficiency increases for contracted basis sets.
- Exploring on-the-fly contraction



Direct solution for the density matrix

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



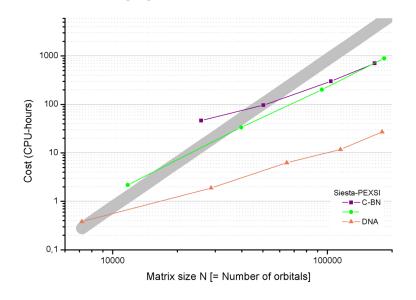


$$\hat{\rho} = Im \left(\sum_{l=1}^{P} \frac{\omega_l}{H - (z_l + \mu)S} \right)$$

For sufficiently big problems (quasi-)1D: $\mathcal{O}(N)$ (quasi-)2D: $\mathcal{O}(N^{3/2})$ 3D: $\mathcal{O}(N^2)$

(Due to sparsity of the target density matrix)

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



Solver strategies for performance and features: Use external libraries

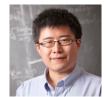
ELSI initiative to integrate solver libraries



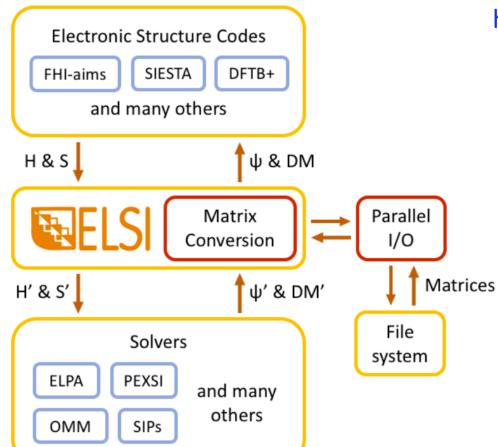
Volker Blum, Duke



Lin Lin, Berkeley



Jiangfen Lu, Duke



https://elsi-interchange.org

Interface in Siesta:

Collaboration with Victor Yu (Duke)





Solver strategies for performance and features: Use external libraries

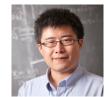
ELSI initiative to integrate solver libraries



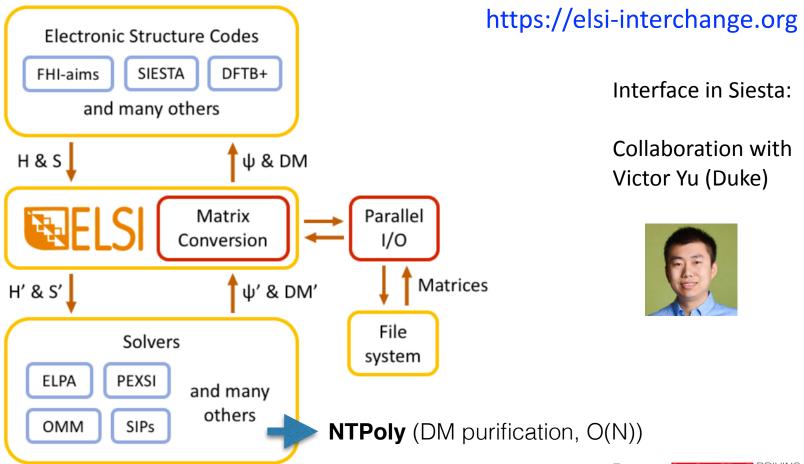
Volker Blum, Duke



Lin Lin, Berkeley

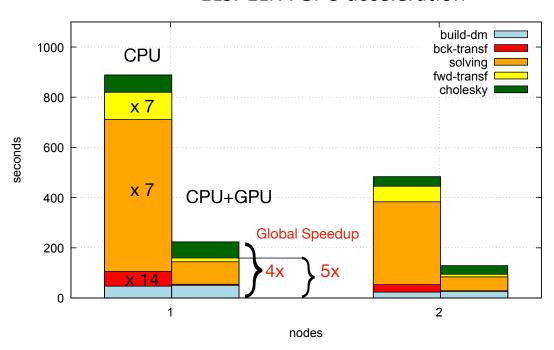


Jiangfen Lu, Duke



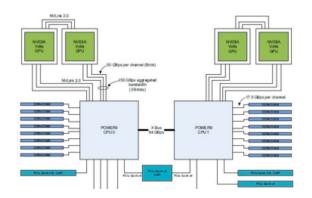
GPU acceleration for diagonalization

ELSI-ELPA GPU acceleration

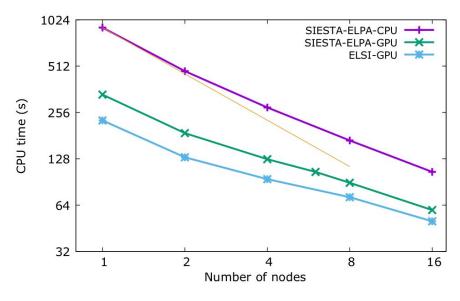


Future enhancements in ELPA (better kernels) and in ELSI (e.g. build-DM stage) are integrated in SIESTA automatically

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



Proper binding of GPUs to MPI ranks



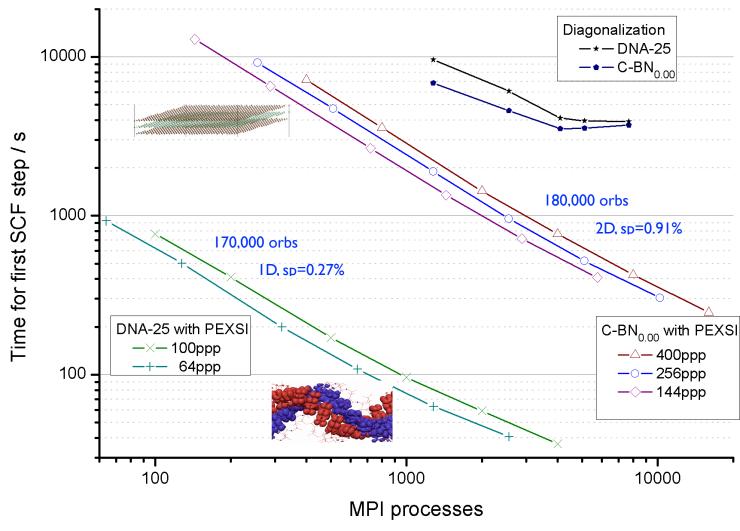


Massive scalability: PEXSI solver

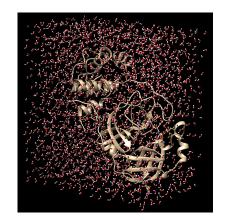
$$\hat{\rho} = Im \left(\sum_{l=1}^{P} \frac{\omega_l}{H - (z_l + \mu)S} \right)$$

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)

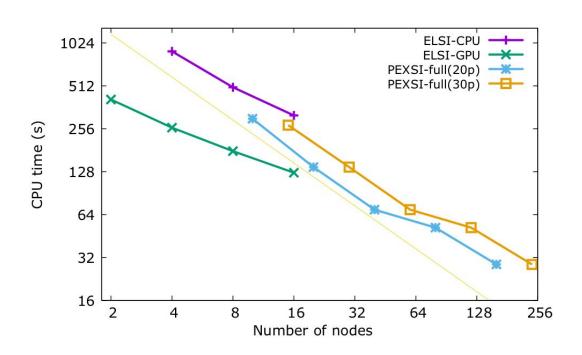


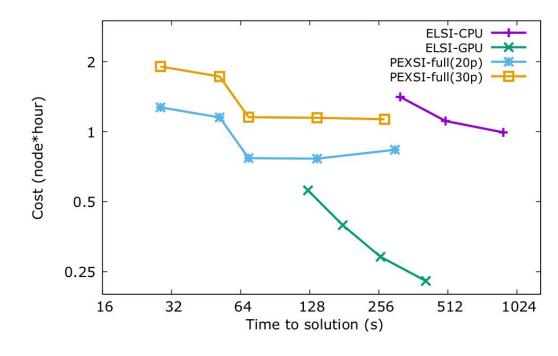
Comparison of global efficiency of solvers for a very large problem



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

Approx 8800 atoms; 58000 orbitals





Work on GPU acceleration of PEXSI library is under way





Follow us on:

THANKS

11



in company/max-centre/



@max_center2



http://www.max-centre.eu/

22/09/2020