

Linear Scaling Solvers for Density Functional Theory Calculations

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Standard Practice

- Choose a basis set $\{\Phi_u\}$ to represent the wavefunctions:
 - $\Psi_i = \sum c_{ui} \Phi_u$.
- Compute the overlap matrix:
 - $S_{uv} = \int \Phi_u^* \Phi_v$.
- Construct the operator matrix:
 - $H_{uv} = \int \Phi_u^* \mathcal{H} \Phi_v$.
- Solve the generalized eigenvalue problem:
 - $H_{uv} C = S_{uv} C \lambda$.
- Compute the density matrix in the chosen basis using the eigenvectors:
 - $P_{uv} = \sum c_{ui} c_{vi}$.
- However, this conventional method requires $O(N^3)$ operations.

$O(N^3)$ Is A Problem

- x10 the number of atoms requires x1000 the computational power.
- Large systems of interest: dilute solutions, defects in a solid, biological molecules, etc.
- Kohn's Nearsightedness principle: Changes of that potential, no matter how large, beyond a distance R have limited effects on local electronic properties.
- Practically speaking: H and P should be sparse, and only grow linearly with the system size.

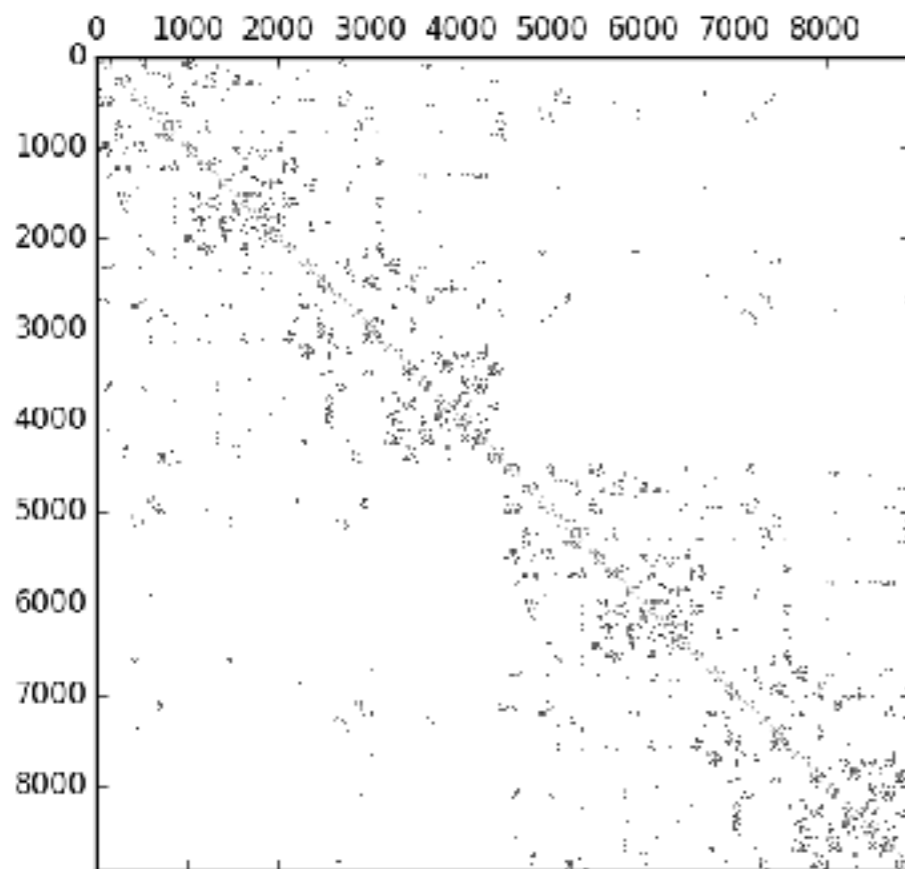
Locality Example

- Maximally Localized Wannier Functions (Boys Orbitals).
- Compute MLWF, and truncate outside of some radius.
- Example: 64 Water Molecules computed with a plane wave code.

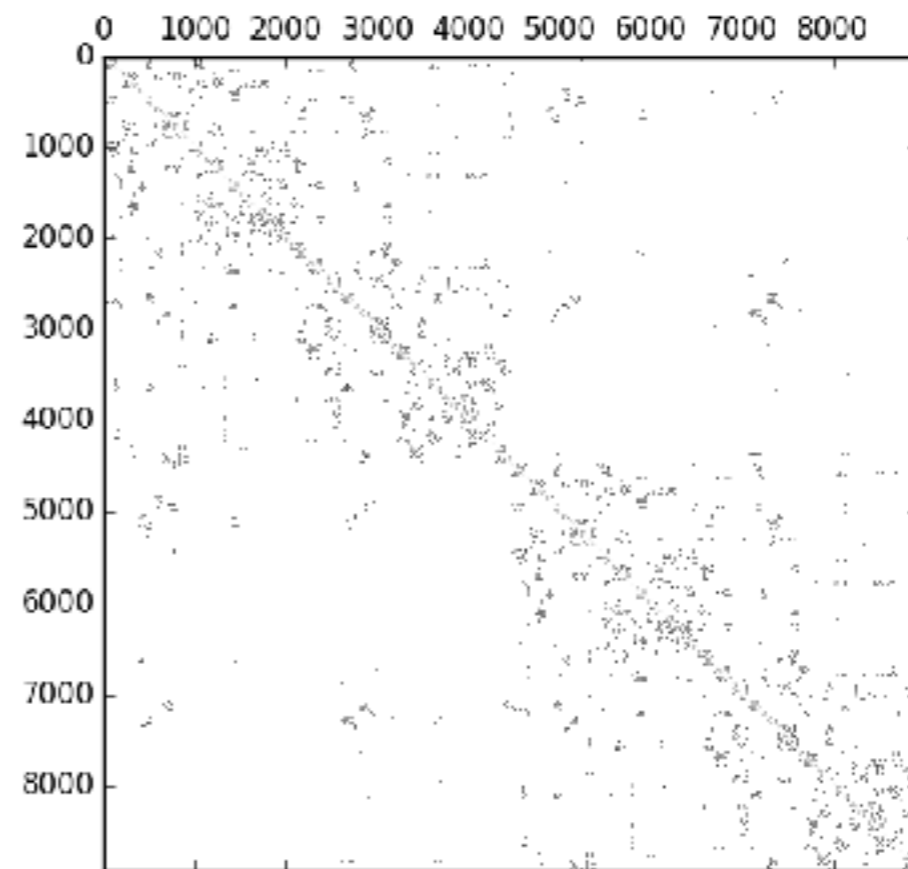
Radius (Bohr Radii)	Relative Energy Error
14	1.93E-06
10	3.46E-05
8	1.90E-04
5	2.52E-03

Locality Example-2

- CP2K linear scaling semi-empirical calculations.
- 2D polymer system, 3168 atoms.



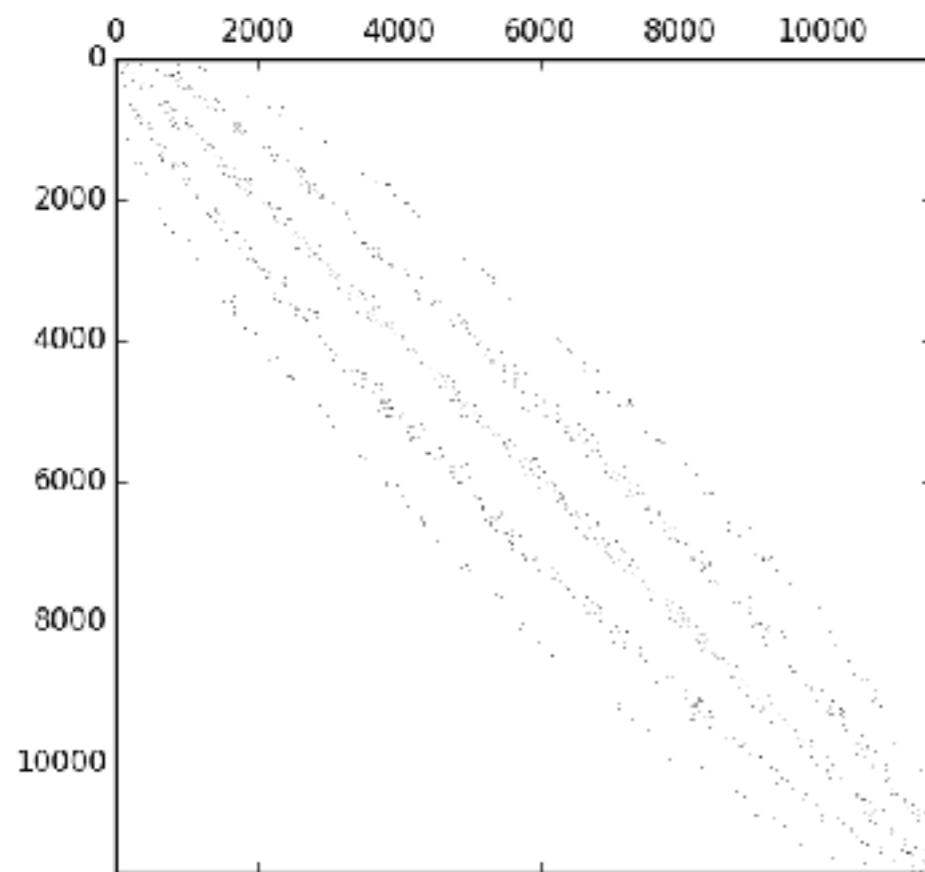
Hamiltonian



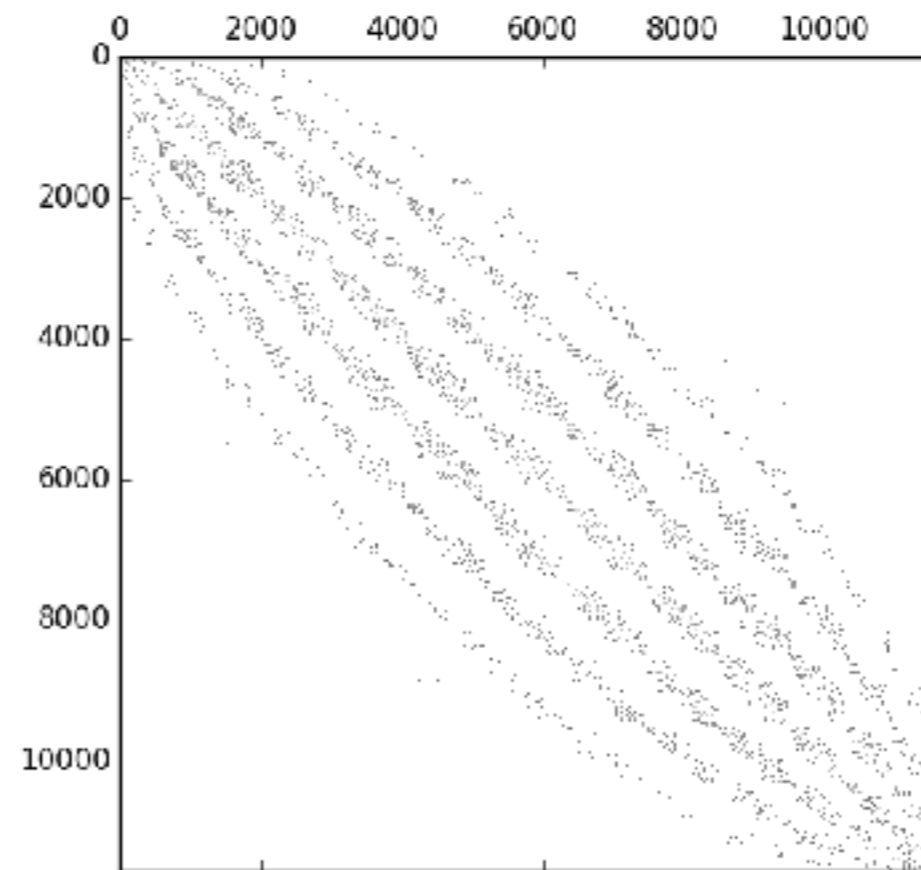
Density Matrix

Locality Example-3

- BigDFT linear scaling calculation.
- 1920 Water Molecule Cluster.



Hamiltonian



Density Matrix

Unified Framework For Solvers

- There exist a variety of linear scaling methods in the literature.
- For the dense eigenvalue problem, there exist a number of massively parallel libraries.
- For sparse eigenvalue problem as well, though aiming for dense eigenvectors.
- “No such standard library” for the sparse algorithms used in linear scaling quantum chemistry.
- Need a unifying mathematical framework that ties these variety of methods all together.

Matrix Functions

- Function of one variable: $f(x)$.
- Function of a matrix: $f(A)$.
- Examples:
 - $f(A) = A^2$.
 - $f(A) = A^{-1}$.
 - $f(A) = e^A$.
- Applications:
 - Differential Equations.
 - Graph Theory.
 - Lattice Quantum Chromodynamics.
 - Of course: Electronic Structure Calculation.

Matrix Function Applications

- Differential Equations:
 - $d^2y/dt^2 + Ay = 0$, $y(0) = y_0$, $y'(0) = y'_0$
 - $y(t) = \cos(\sqrt{A}t)y_0 + \sqrt{A}^{-1}\sin(\sqrt{A}t)y'_0$.
- Graph Theory:
 - Let A be the adjacency matrix.
 - The ease of sending a message from node i to j : $e^{A_{ij}}$.
 - The Katz centrality of a node: $(I-cA)^{-1}_{ii}$.

Electronic Structure Applications-1

- Basis Set Orthogonalization (Lowdin):
 - Matrix inverse square root function.
 - $HC = SC\lambda$.
 - $S^{-1/2}HS^{-1/2}C = C\lambda$.
- Overlap Matrix Inversion (Orbital Minimization Method):

$$\begin{aligned} E_{KS}[\{\phi_i\}_{i=1}^N] &= \sum_{i,j=1}^N (S^{-1})_{ij} \int_{\Omega} \phi_i(\mathbf{r}) \left(-\frac{1}{2}\Delta \right) \phi_j(\mathbf{r}) d\mathbf{r} \\ &+ \frac{1}{2} \int_{\Omega} \int_{\Omega} \frac{\rho(\mathbf{r}_1)\rho(\mathbf{r}_2)}{|\mathbf{r}_1 - \mathbf{r}_2|} d\mathbf{r}_1 d\mathbf{r}_2 + E_{XC}[\rho] \\ &+ \sum_{i,j=1}^N (S^{-1})_{ij} \int_{\Omega} \phi_i(\mathbf{r}) [V_{ext}\phi_j](\mathbf{r}) d\mathbf{r}, \end{aligned}$$

Electronic Structure Applications-2

- Curvy Step Density Matrix Minimization:

- $P_{n+1} = e^{-X}P_n e^X.$

- McWeeny Purification:

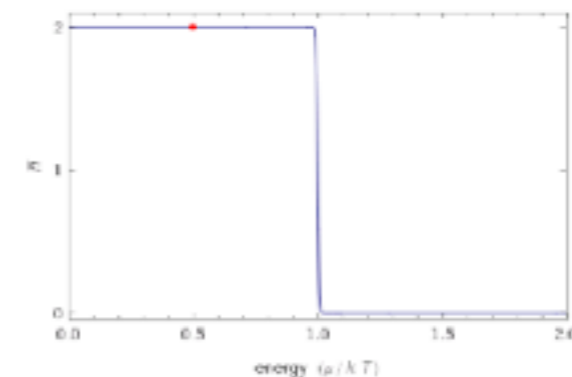
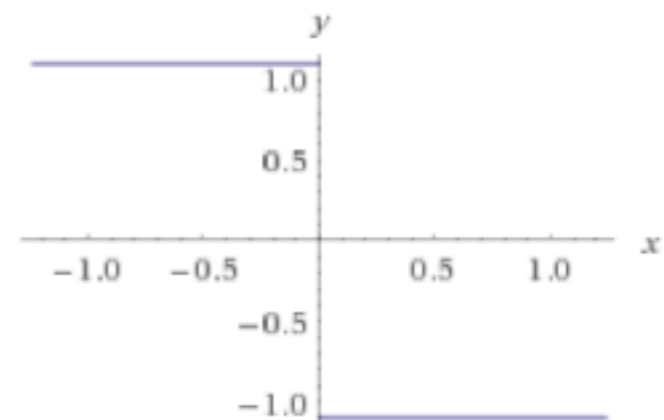
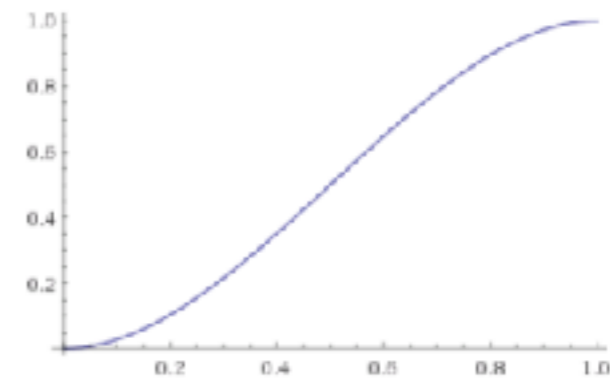
- $P_{n+1} = 3P_n^2 - 2P_n^3.$

- Matrix Sign Function:

- $P = 0.5*(I - \text{sign}(H - \mu I)).$

- Fermi Operator Expansion:

- $P = 0.5*(I - \text{erf}((H - \mu I)/(\Delta\varepsilon))).$



How To Compute A Matrix Function

- If The Matrix Is Diagonalizable:
 - $A = UDU^*$.
 - $f(A) = Uf(D)U^*$.
 - Looks familiar...
- If It Is Not:
 - A variety of approaches, including Schur Decomposition based approach:
 - $A = UTU^*$.
 - $f(A) = Uf(T)U^*$. Explicit formulas for $f(T)$ exists.

How To Compute A Matrix Function-2

- Those direct methods are equally (or more) expensive than our original approach!
- Instead, we can approximate our function with polynomials based on Chebyshev Polynomials, truncated Taylor Series, etc...
- For example, Taylor series approximation of a matrix exponential:
 - $e^A = I + A + A^2/2! + A^3/3! + \dots$
- Computing a matrix polynomial only requires matrix multiplication, addition.
- When the matrices are sparse, we can use sparse matrix multiplication and addition, which scale linearly with the system size.

Feature Set Summary

- **General Polynomials**

- Standard Polynomials
- Chebyshev Polynomials

- **Transcendental Functions**

- Trigonometric Functions
- Exponential and Logarithm

- **Matrix Roots**

- Square Root and Inverse Square Root
- Arbitrary Rational Roots

- **Quantum Chemistry**

- Density Matrix Minimization
- Density Matrix Purification
- Chemical Potential

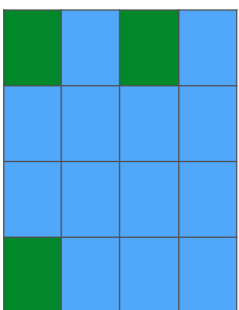
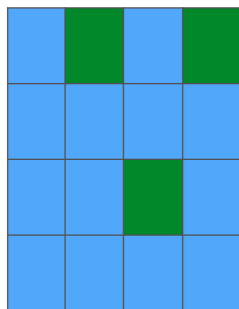
- **Other**

- Matrix Inverse
- Sign Function
- Load Balancing Matrices
- File I/O

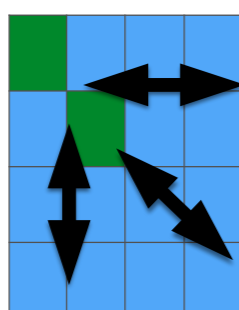
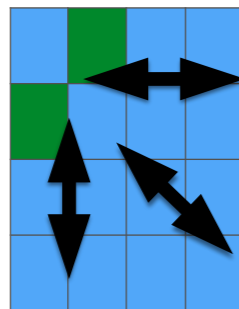
Library Integration - 1

- Challenge: Integrate With Parallel Programs Using a Variety of Different Data Layouts.
- Solution 1: Parallel File I/O through the standard matrix market format for rapid prototyping.
- Solution 2: Arbitrary Data Remapping Routines.

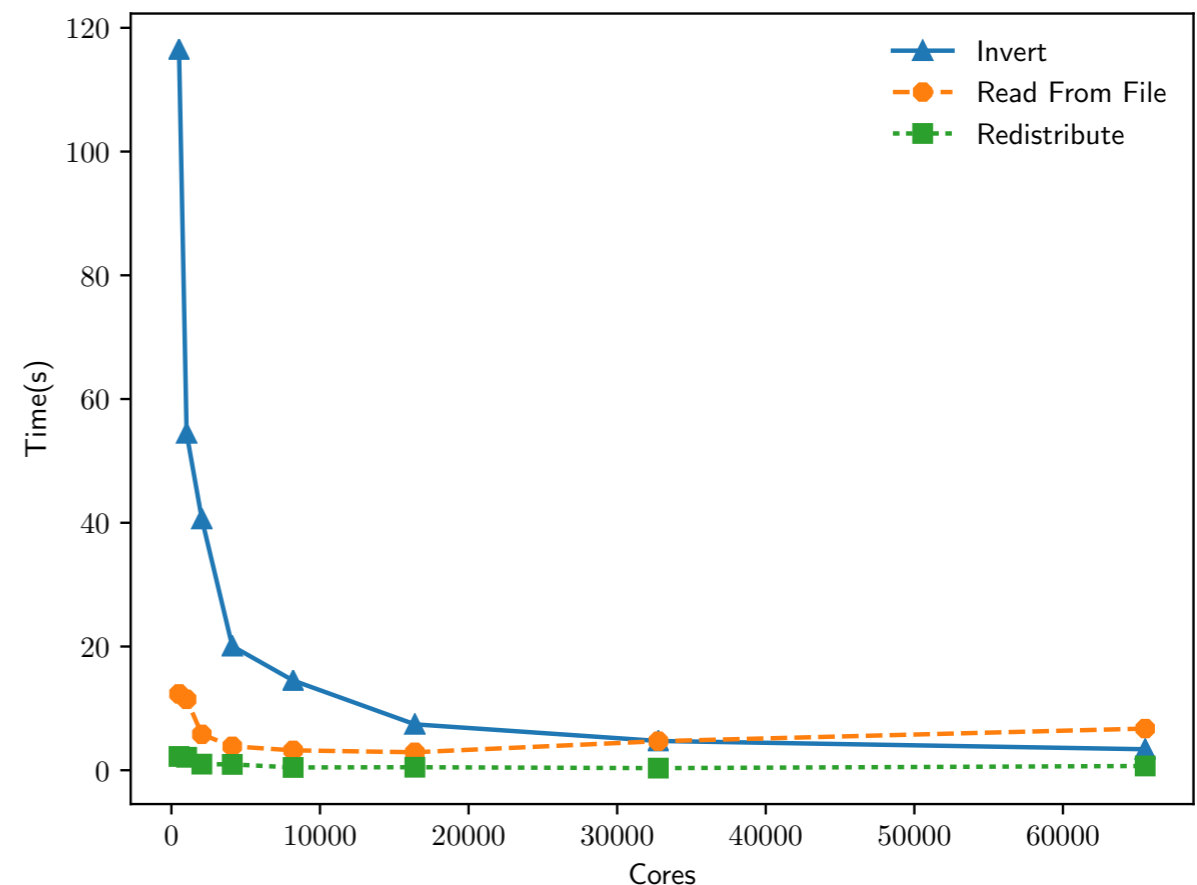
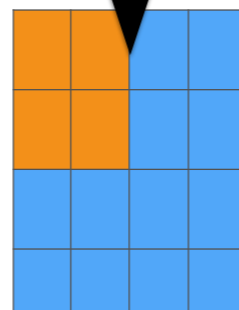
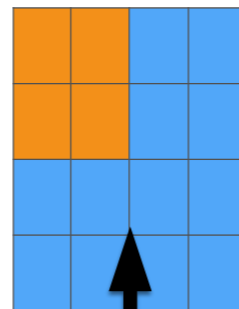
Start With Arbitrary Elements



2D All-To-All Gather

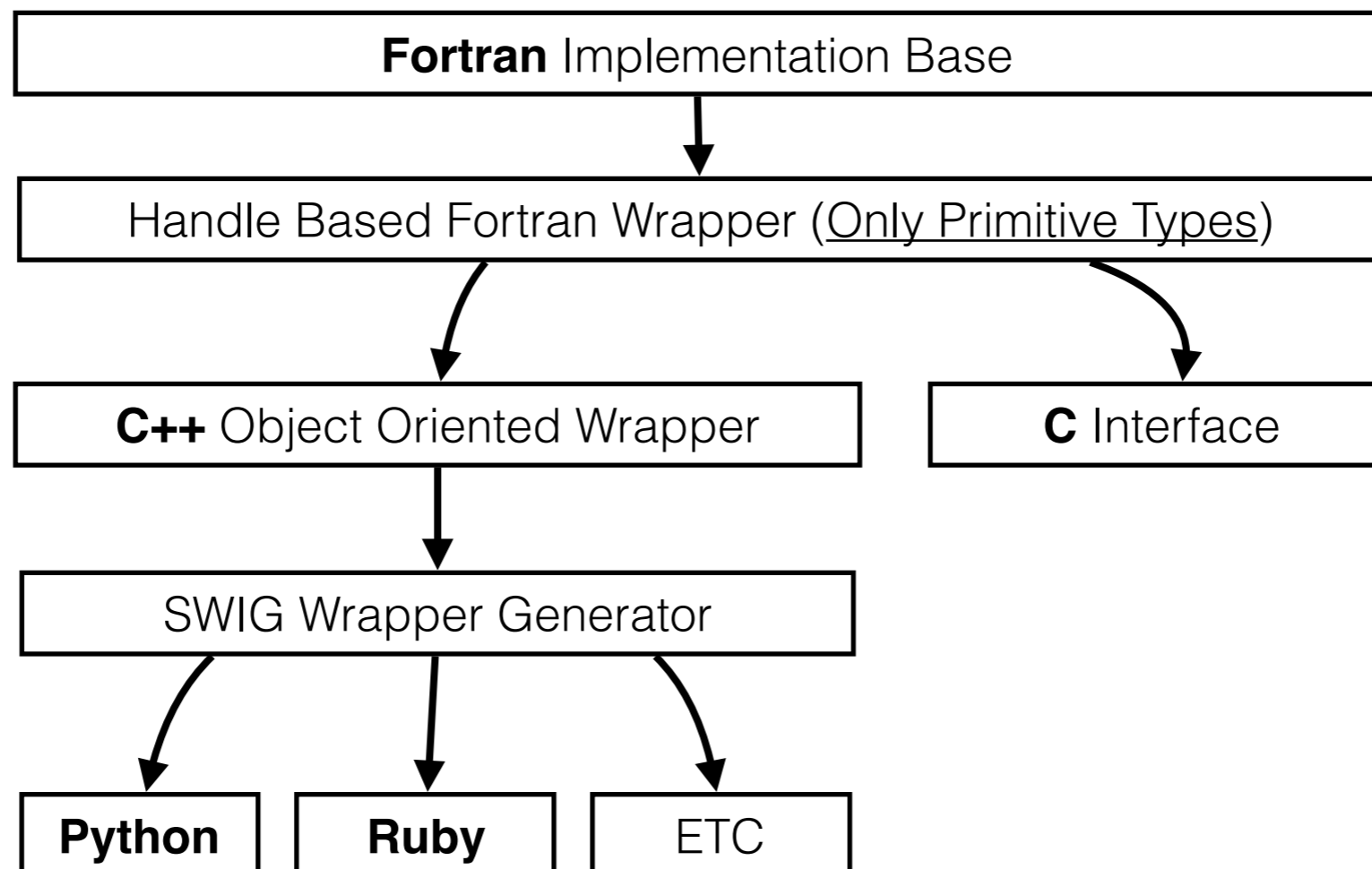


Sum Blocks



Library Integration - 2

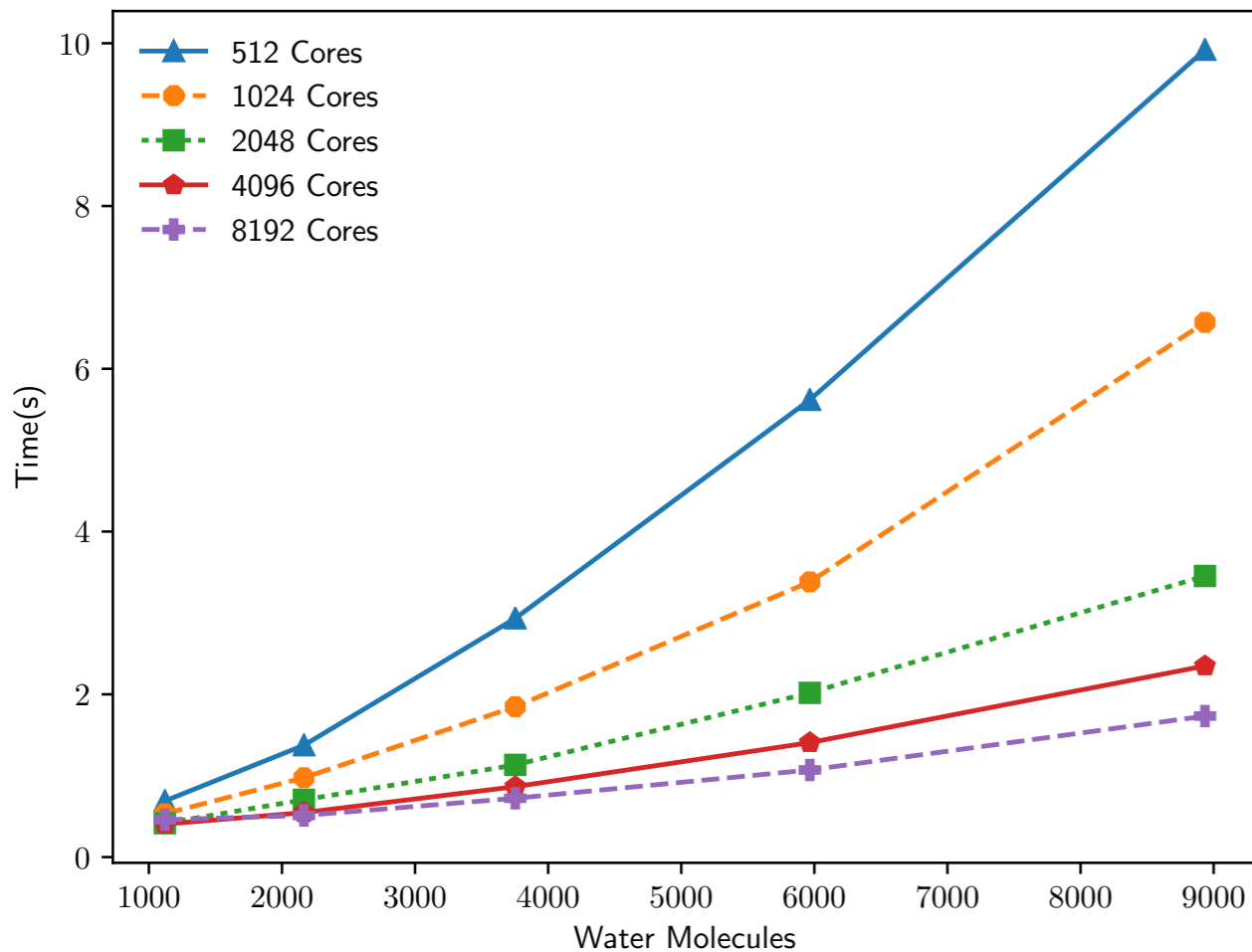
- Challenge: Integration with codes written in a variety of programming language.
- Solution: Programming Language Wrapper Hierarchy.



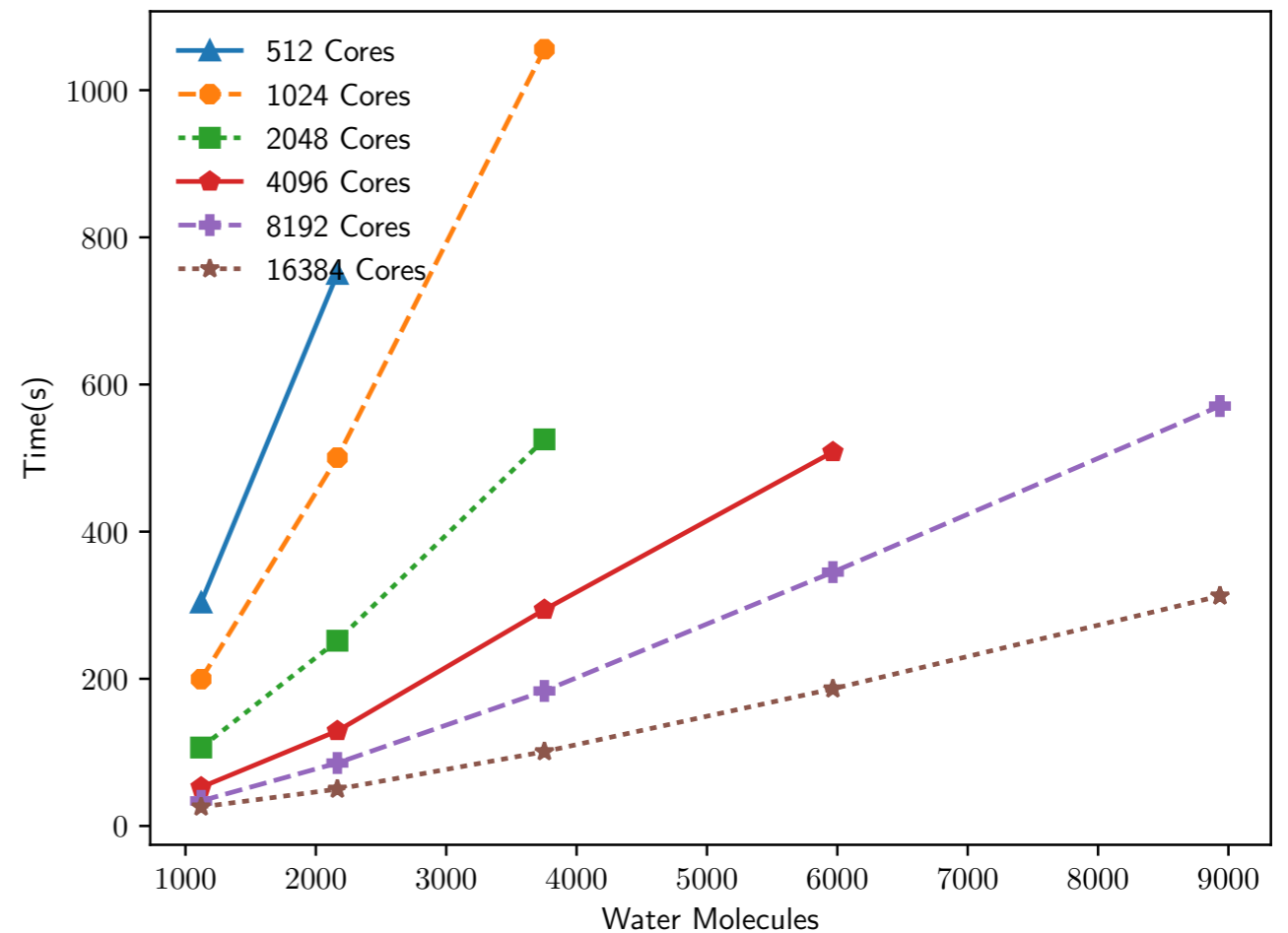
Application - 1

- Compute The Inverse Square Root of the Overlap Matrix.
- System: Water Clusters.

STO Basis Set



DZP Basis Set



Application - 2

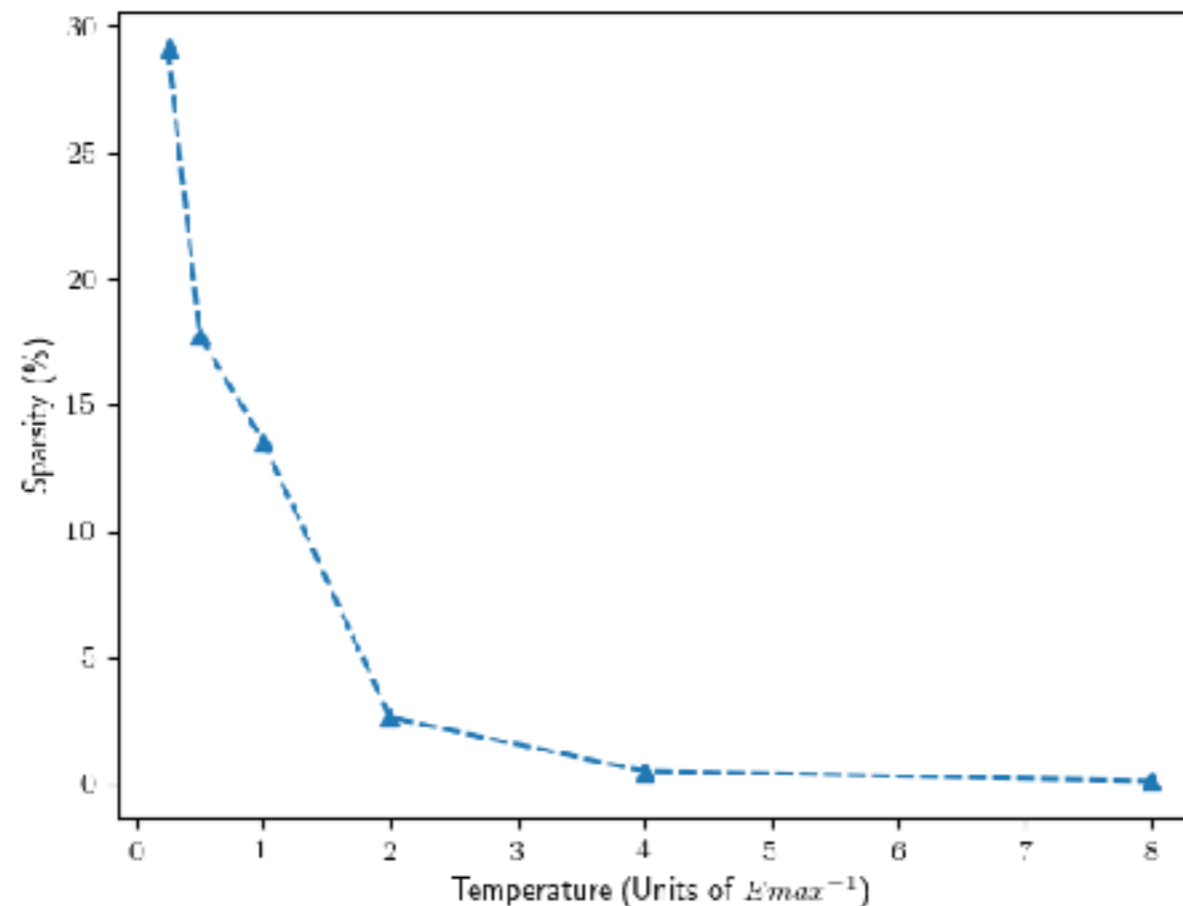
- Comparison of Density Matrix Solver Methods.
- Programs Compared: ErgoSCF, CP2K, BigDFT.
- Basis Sets: Gaussians, Gaussians, Wavelets.
- Theories: Hartree-Fock, Tight Binding, DFT (PBE).
- Solver Methods: TRS2, TRS4, HPCP.

	ErgoSCF	BigDFT	CP2K
Basis Functions	13,468	11,544	11,544
Sparsity of $S^{-\frac{1}{2}}HS^{-\frac{1}{2}}$	2.88%	1.36%	2.75%
Sparsity of P	7.36%	4.11%	5.85%
Occupation	71.4%	66.7%	66.7%

Method	ErgoSCF (s)	BigDFT (s)	CP2K (s)
TRS2	0.73	0.78	0.50
TRS4	0.71	0.78	0.43
HPCP	1.09	0.91	0.71

Application - 4

- Data Set: Israeli Social Network “TheMarker Cafe”
- Nodes: 69413. Sparsity: 0.04%.
- Artificially made symmetric.
- How does the ease of communicating with nearest neighbors affect the flow of information on the whole network?
- Estrada’s Scaled Matrix Exponential Metric: $e^{\beta A}$



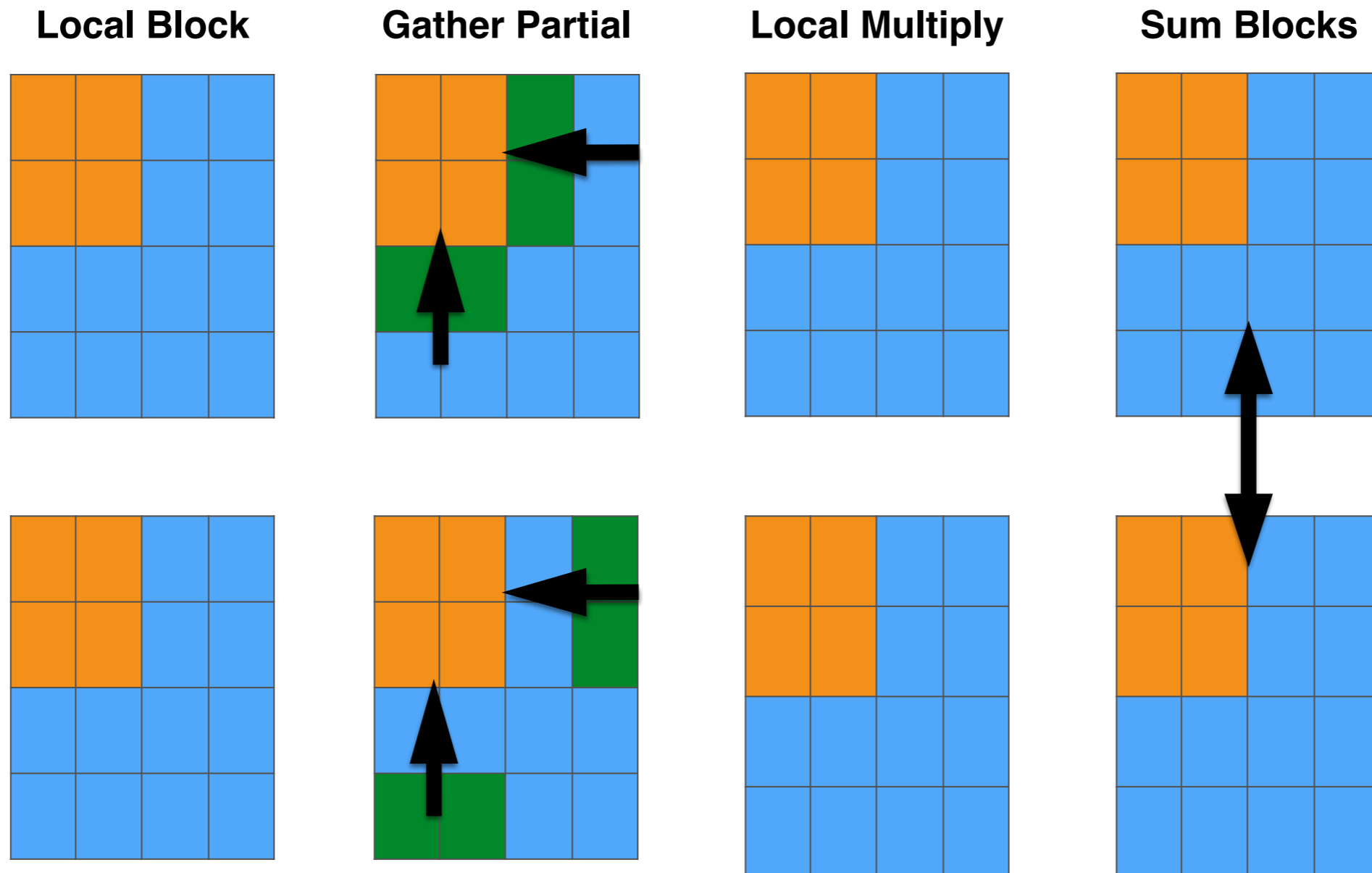
Conclusion



- Massively parallel library for computing the functions of sparse matrices.
- All functions are computed within one unified framework, and optimized by a single kernel (Sparse Matrix-Matrix Multiplication).
- Easy integration into a variety of different programs.
- Promising application areas well beyond quantum chemistry.
- Code availability: <https://github.com/william-dawson/NTPoly>

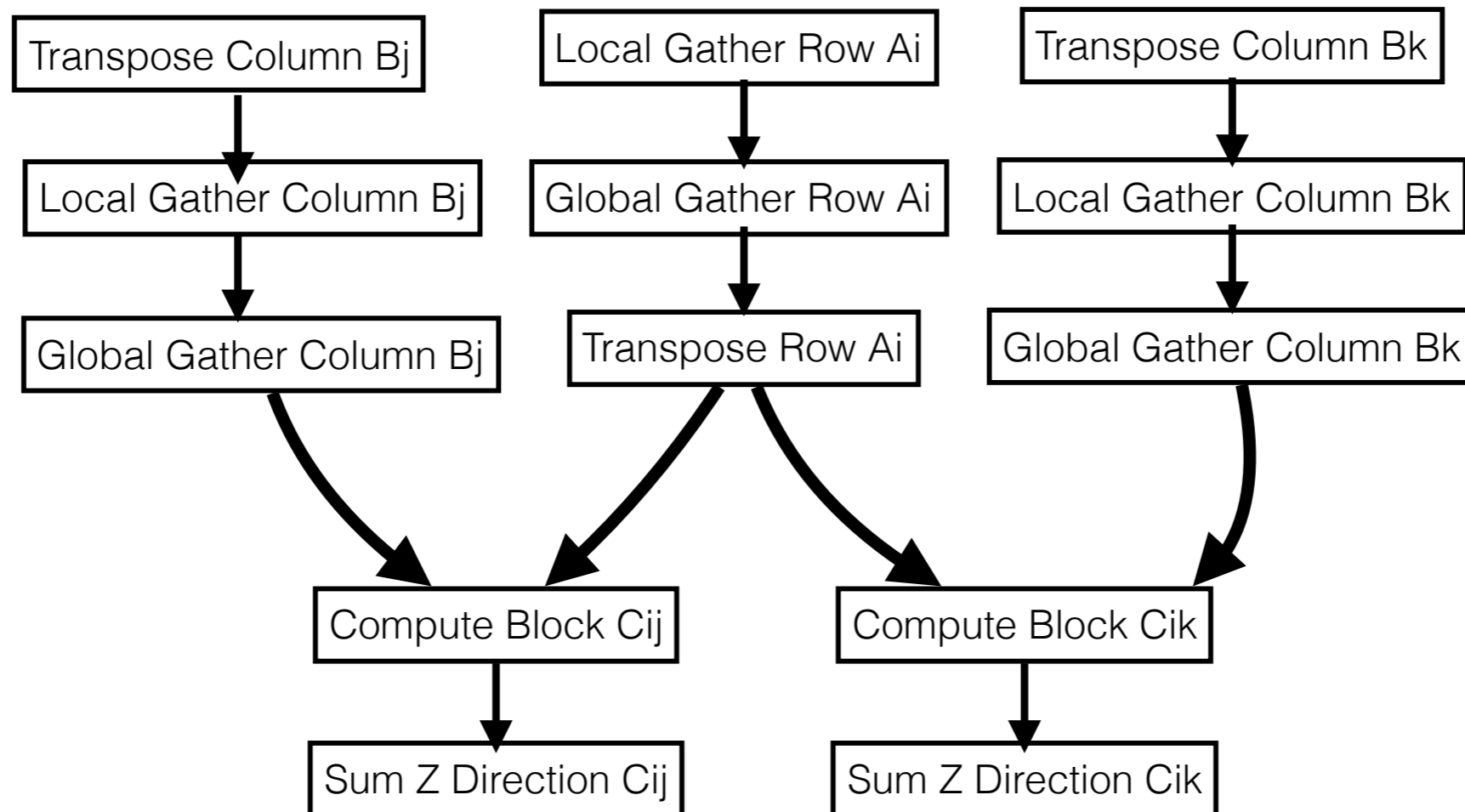
Bonus Slide - Distributed Matrix Multiply

- Based on a communication avoiding algorithm.



Bonus Slide - OpenMP Parallelism

- OpenMP Task Parallelism allows dynamic load balancing, communication/computation overlap.



Bonus Slide - Parallel Performance

- Compute the inverse of a nearly banded matrix.
- 32768×32768 matrix, 1% filled.

