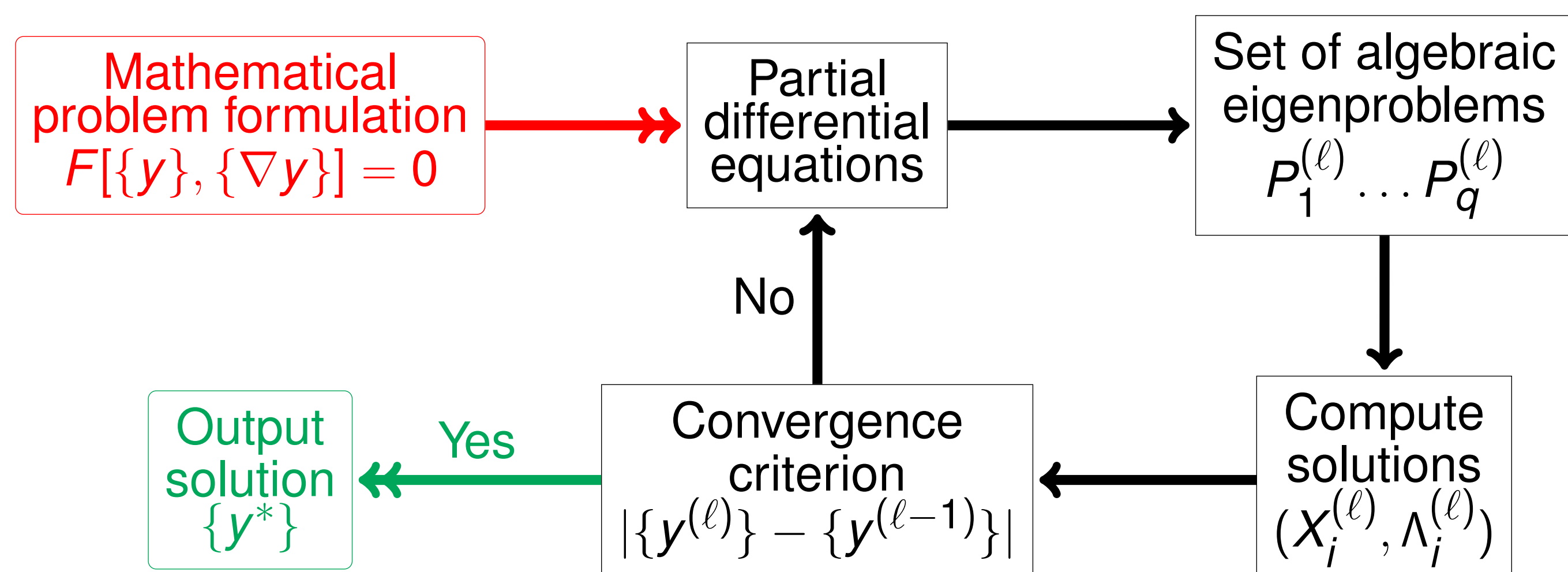


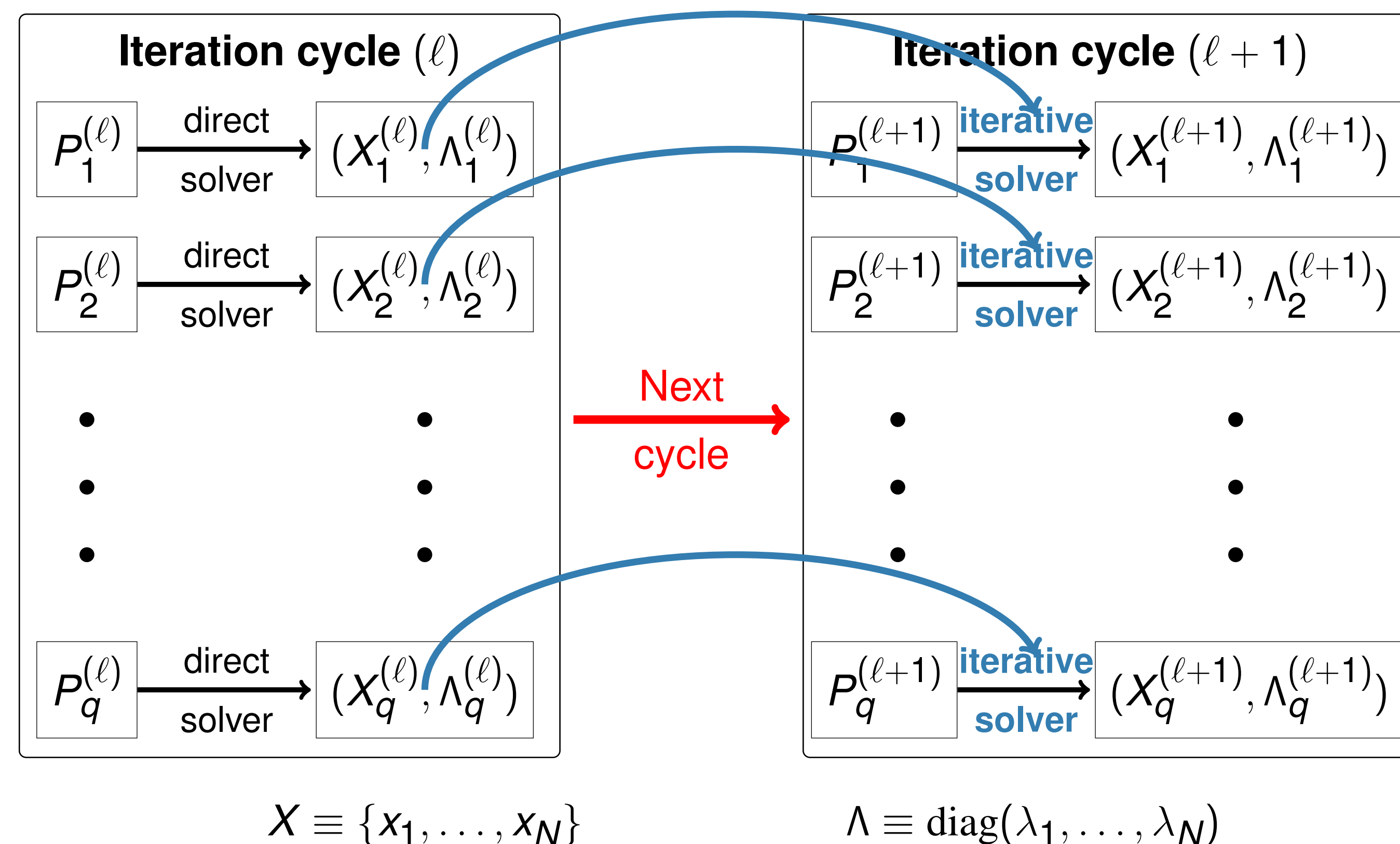
Sequences of Eigenvalue problems



Traditional solving strategy

- Every set $P^{(1)} \dots P^{(\ell)} P^{(\ell+1)} \dots P^{(M)}$ is handled as a set of M disjoint eigenproblems with overall complexity $\sim M \times N^3$;
- Each problem $P^{(\ell)}$ is solved independently using a direct solver as a black-box from a standard library (e.g. ScaLAPACK).

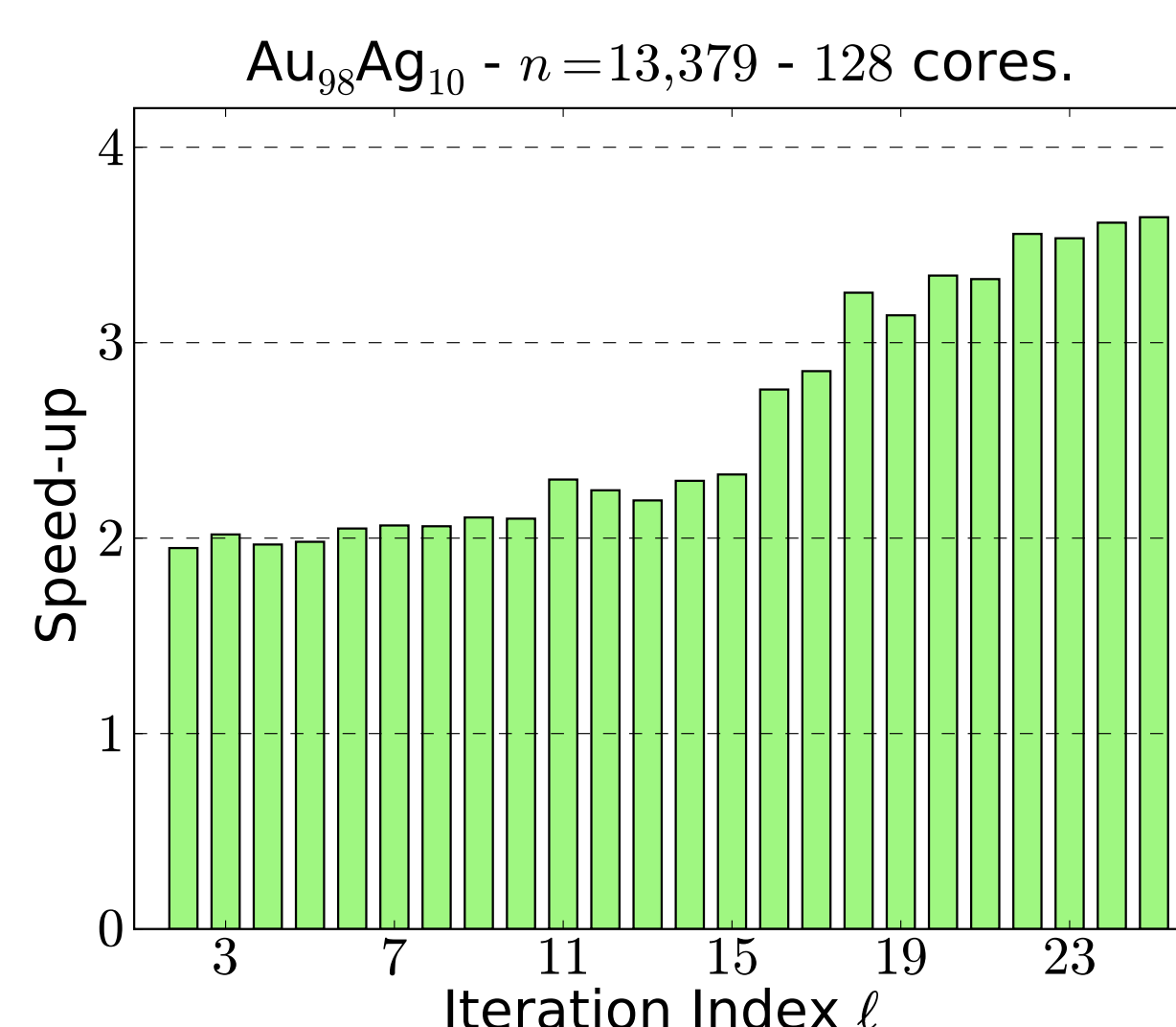
Knowledge oriented solving strategy



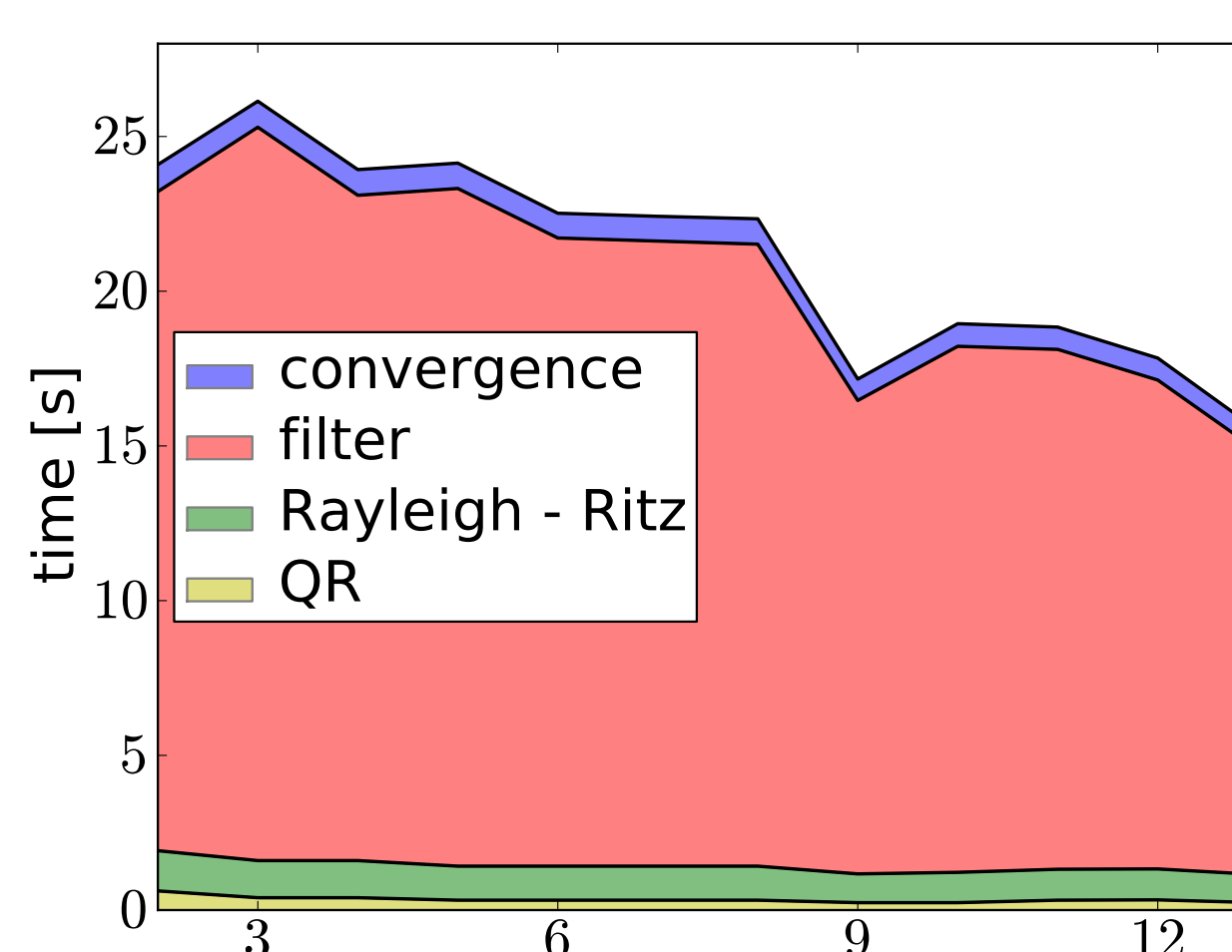
Algorithm 1 Chebyshev Accelerated Subspace Eigensolver

Input: Matrix $A^{(\ell+1)}$, vector matrix $X^{(\ell)}$, values $\Lambda^{(\ell)}$, eigenpairs nev , tolerance tol , poly degree deg , search space nex , parameter opt .
Output: nev extremal eigenpairs $(\Lambda^{(\ell+1)}, X^{(\ell+1)})$.

- 1: Initialize $V = X^{(\ell)}$, $\mu_1 = \lambda_1^{(\ell)}$, $\mu_{\text{nev}+\text{nex}} = \lambda_{\text{nev}+\text{nex}}^{(\ell)}$ ▷ WORKING VARIABLES
- 2: $(b_{\text{sup}} > \lambda_N) \leftarrow \text{LANCZOS}(A^{(\ell+1)})$ ▷ SPECTRAL BOUND
- 3: Set the degrees $m = (m_1, \dots, m_{\text{nev}}) = (\text{deg}, \dots, \text{deg})$.
- 4: **while** $\text{nconv} = \text{size}(X^{(\ell+1)}) < \text{nev}$ **do**
- 5: $V \leftarrow \text{FILTER}(A^{(\ell+1)}, b_{\text{sup}}, \mu_1, \mu_{\text{nev}+\text{nex}}, V, m)$ ▷ USES ARRAY OF DEGREES
- 6: Reorthonormalize $[X^{(\ell+1)} V] = QR$ ▷ QR ALGORITHM
- 7: $Q \leftarrow [Q_{:, \text{nconv}+1} \dots Q_{:, \text{nev}+\text{nex}}]$ ▷ REDUCE TO ACTIVE SUBSPACE
- 8: Compute Rayleigh quotient $G = Q^T A^{(\ell+1)} Q$ ▷ RAYLEIGH-RITZ (Start)
- 9: Solve the reduced problem $GW = W\tilde{\Lambda}$ ▷ DIRECT SOLVER
- 10: Compute $V = QW$ ▷ RAYLEIGH-RITZ (End)
- 11: Lock converged eigenpairs into $(\Lambda^{(\ell+1)}, X^{(\ell+1)})$ ▷ LOCKING
- 12: Reducing (search space) vector matrix V ▷ DEFLATION
- 13: Update variables $\mu_1 = \tilde{\lambda}_1$, $\mu_{\text{nev}+\text{nex}} = \tilde{\lambda}_{\text{nev}+\text{nex}}$
- 14: Compute the degrees $(m_1, \dots, m_a) = m$ ▷ OPTIMIZATION
- 15: **end while**



Above: Speed-up after pre-conditioning
Below: ChASE time profile.



Algorithm 2 Optimized Chebyshev Filter

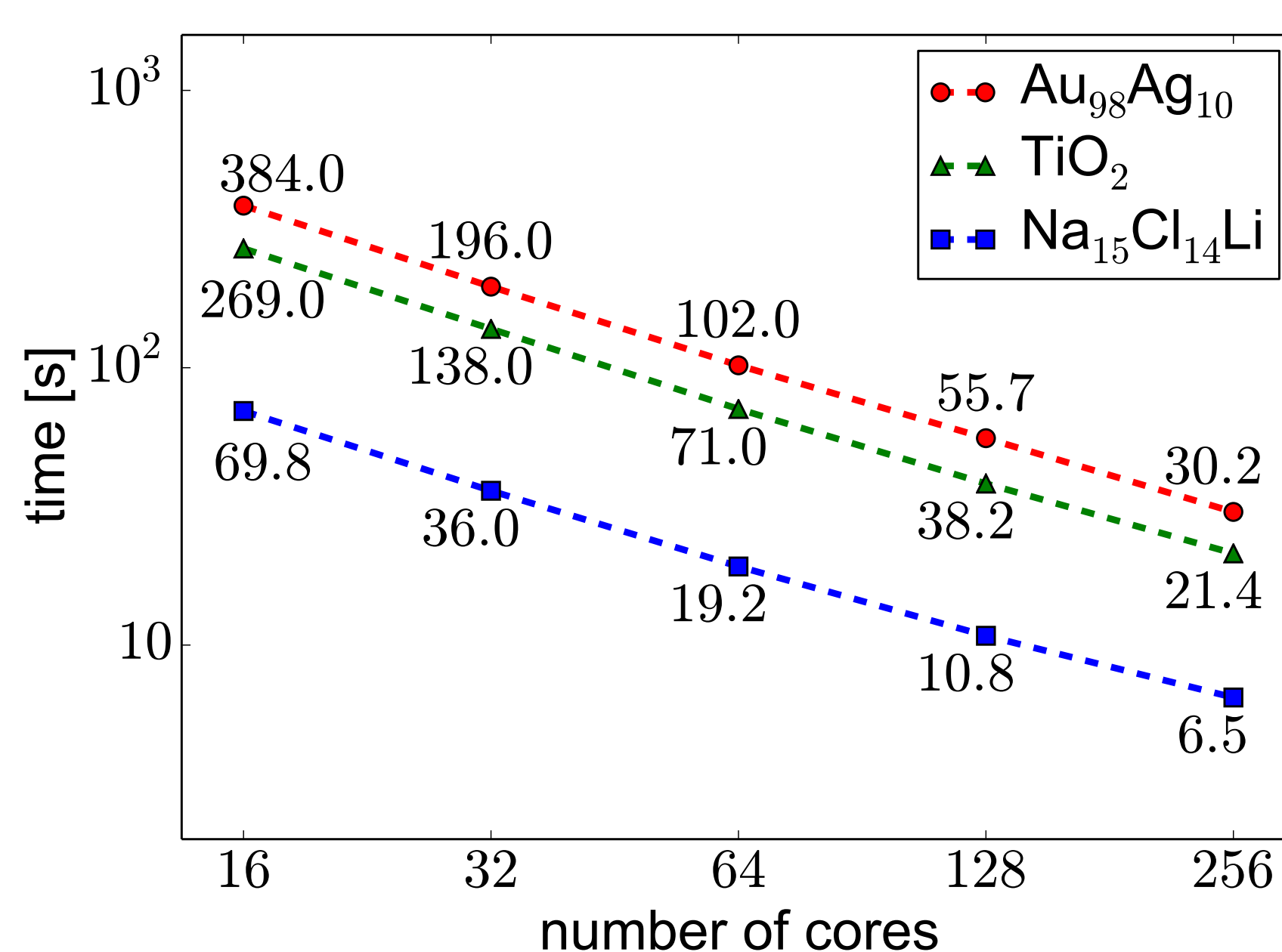
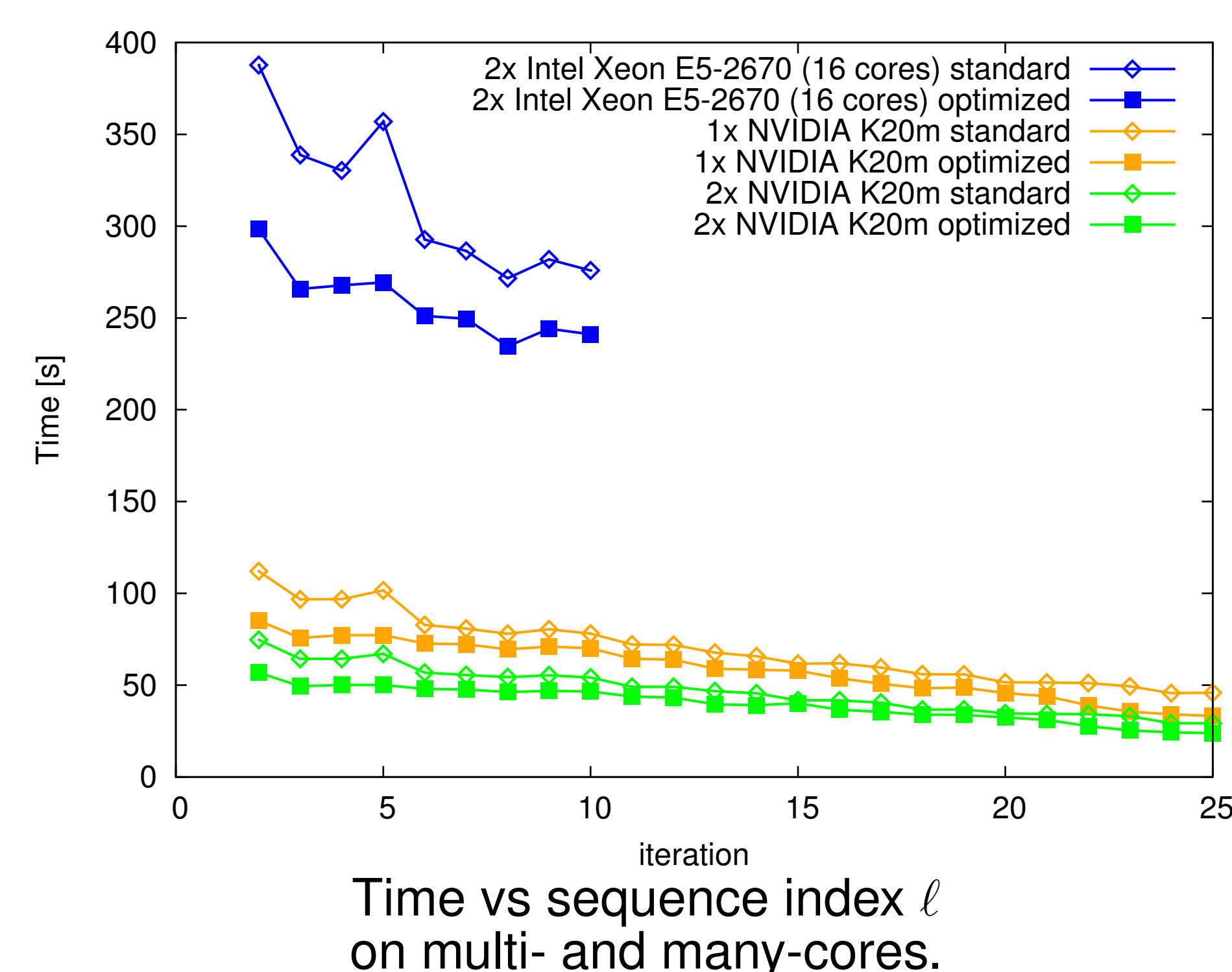
- 1: **procedure** $\text{FILTER}(A^{(\ell+1)}, b_{\text{sup}}, \mu_1, \mu_{\text{nev}+\text{nex}}, V, m = \{m_1, \dots, m_k\})$
- 2: $c = \frac{b_{\text{sup}} + \mu_{\text{nev}+\text{nex}}}{2}$; $e = \frac{b_{\text{sup}} - \mu_{\text{nev}+\text{nex}}}{2}$;
- 3: $\sigma_1 = \frac{e}{\mu_1 - c}$;
- 4: $V \leftarrow \frac{\sigma_1}{e} (A - cI_n) V$;
- 5: $s \leftarrow \text{argmin}_{a=1, \dots, k} m_a \neq 1$
- 6: **for** $i = 1, \dots, m_k - 1$ **do**
- 7: $\sigma_{i+1} \leftarrow \frac{1}{\frac{2}{\sigma_1} - \sigma_i}$
- 8: $V_{s:n} \leftarrow Z_{s:n}$
- 9: $Z_{s:n} \leftarrow 2 \frac{\sigma_{i+1}}{e} (A - cI_n) V_{s:n} - \sigma_{i+1} \sigma_i V_{s:n}$
- 10: $s \leftarrow \text{argmin}_{a=s, \dots, k} m_a \neq i + 1$
- 11: **end for**
- 12: **return** $V \leftarrow Z$
- 13: **end procedure**

$$Z \leftarrow 2 \frac{\sigma_{i+1}}{e} (A - cI) \times V - \sigma_{i+1} \sigma_i V$$

xGEMM

C++ implementations of ChASE

- Multi-core: Multi-threaded BLAS and OpenMP;
- Distributed: MPI based on the `Elemental` library;
- Many-core: CUDA for one or multiple GPUs;
- Distributed CPU/GPU — work in progress.



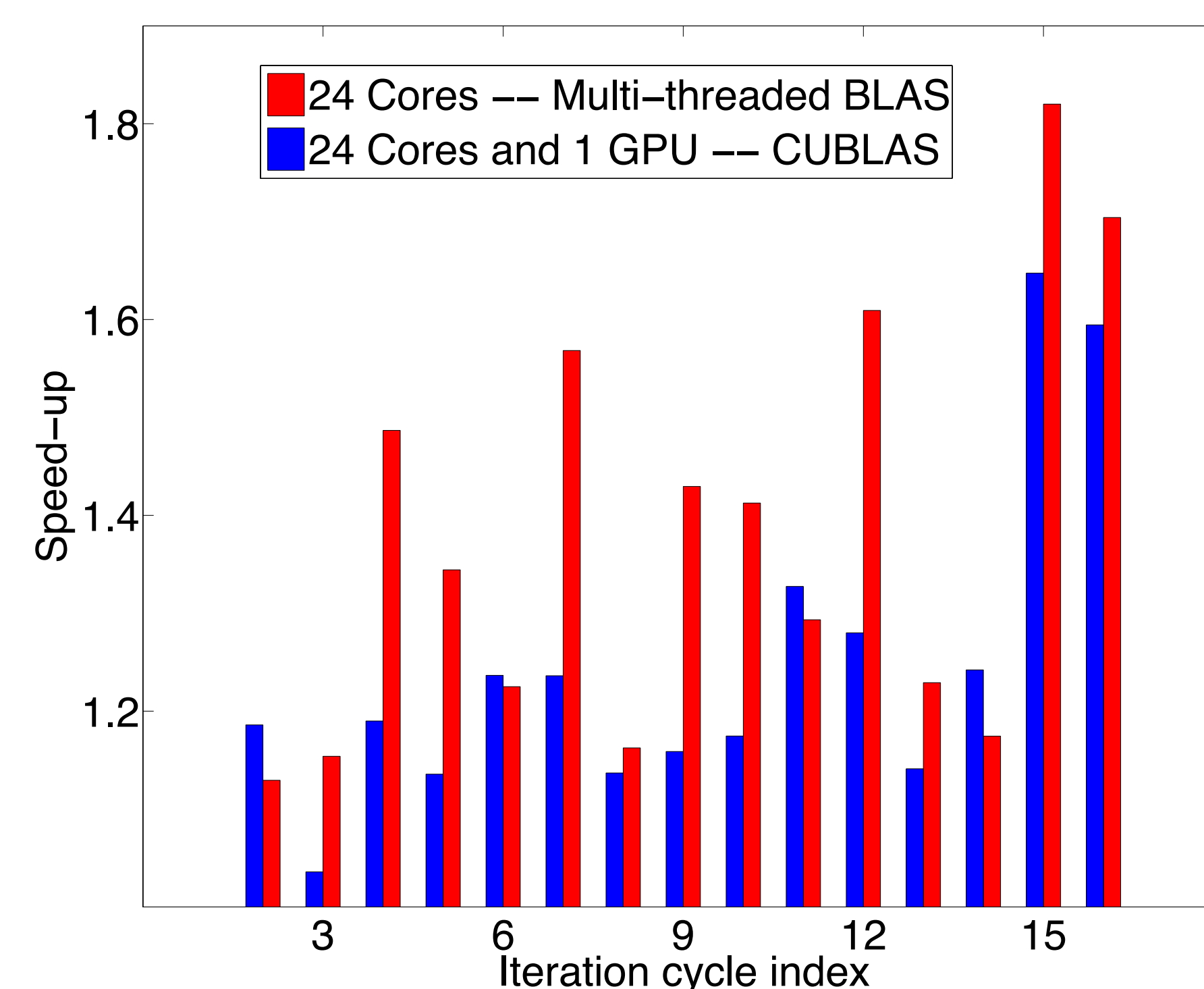
Strong scalability of the MPI implementation
System sizes: $N = 13,379 - 12,455 - 9,273$.

The ChASE library

- Release date: 30th of June 2016;
- Documentation: generated through Sphinx;
- License: To Be Determined.

An array of optimal polynomial degrees

Computed using convergence ratio estimates, it minimizes `mat-vec` multiplications.



Speed-up after degree optimization.