Parallel solution of the generalized eigenvalue problem given in a factored form

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Outline of the talk:

- description of the problem,
- brief description of the two-sided Hari–Zimmermann (HZ) algorithm for the GEP,

- implementation details of the parallel algorithm,
- partial results of numerical testing.

DFT – Density Functional Theory

Density Functional Theory framework

- is used in simulation of the physical properties of complex quantum mechanical systems made of few dozens up to few hundreds of atoms
- the core of the method relies on the simultaneous solution of a set of Schrödinger–like equations also known as Kohn–Sham equations
- there exists a wide variety of approaches that can be used to "translate" the DFT mathematical layout into a computational tool.

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Full-potential Linearized Augmented Plane Wave (FLAPW) method

- FLAPW method is one od the most accurate methods particular discretization of the DFT fundamental equations
- FLAPW is all-electron method it explicitly describes all of the (potentially large number of) electrons in the material with a much larger number of basis function

• it is a quite computationally expensive method.

FLAPW method

Full-potential Linearized Augmented Plane Wave (FLAPW) method

the discretization in FLAPW method leads to the solution of the generalized eigenvalue problem for matrices (H, S), where

$$H = \sum_{a=1}^{N_A} (A_a^* T^{[AA]} A_a + A_a^* T^{[AB]} B_a + B_a^* T^{[BA]} A_a + B_a^* T^{[BB]} B_a)$$
$$S = \sum_{a=1}^{N_A} (A_a^* A_a + B_a^* U_a^* U_a B_a),$$

where $A_a, B_a \in \mathbb{C}^{N_L \times N_G}$, $T_a^{[\cdots]} \in \mathbb{C}^{N_L \times N_L}$, $U \in \mathbb{C}^{N_L \times N_L}$ is a diagonal matrix, while $(T^{[AA]})^* = T^{[AA]}$, $(T^{[BB]})^* = T^{[BB]}$, and $(T^{[AB]})^* = T^{[BA]}$.

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Typical matrix sizes

- ▶ $N_A = O(100), N_G = O(1000) O(10000), \text{ and } N_L = O(100)$
- ▶ test examples NaCl N_A = 512, N_L = 49
 ▶ N_G = 2256, N_G = 3893, N_G = 6217, N_G = 9273
- test examples $AuAg N_A = 128$, $N_L = 121$
 - ▶ $N_G = 3275$, $N_G = 5638$, $N_G = 8970$, $N_G = 13379$.

Computation of H and S

Proposed by Fabregat-Traver at al.

• write *H* as $H = H_{AA} + H_{AB+BA+BB}$

$$H_{AA} = \sum_{a=1}^{N_A} A_a^* T^{[AA]} A_a$$
$$H_{AB+BA+BB} = \sum_{a=1}^{N_A} \left(B_a^* T^{[BA]} A_a + A_a^* T^{[AB]} B_a + B_a^* T^{[BB]} B_a \right)$$
$$= \sum_{a=1}^{N_A} \left(B_a^* Z_a + Z_a^* B_a \right)$$

(ZHER2Ks!), where

$$Z_a = T^{[BA]}A_a + \frac{1}{2}T^{[BB]}B_a.$$

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Modification?

Why

- the algorithm proposed by Fabregat–Traver at al. computes in parallel only *H* and *S* — then use any GEVD,
- intention to keep matrices in a factored form ideal for parallelization of the GEVD
- usage of one-sided methods faster than the two-sided methods — columnwise action
- such approach usually computes small eigenvalues more accurately
- similar algorithm for the (real) generalized SVD is approximately 125 times faster than the LAPACK routine with theaded MKL.

Transform the problem!

Transformed problem

▶ by using the properties of matrices T^[…] it is obvious that the problem can be written as

$$H = \sum_{a=1}^{N_A} \begin{bmatrix} A_a^* & B_a^* \end{bmatrix} \begin{bmatrix} T^{[AA]} & T^{[AB]} \\ (T^{[AB]})^* & T^{[BB]} \end{bmatrix} \begin{bmatrix} A_a \\ B_a \end{bmatrix} := \sum_{k=1}^n H_k^* T_k H_k,$$
$$S = \sum_{a=1}^{N_A} \begin{bmatrix} A_a \\ B_a \end{bmatrix} U_a^* \begin{bmatrix} A_a \\ U_a B_a \end{bmatrix} := \sum_{k=1}^n S_k^* S_k.$$

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Transform the problem!

... or as products of three (two) matices

$$H = \begin{bmatrix} H_1^* & \cdots & H_n^* \end{bmatrix} \begin{bmatrix} T_1 & & \\ & \ddots & \\ & & T_n \end{bmatrix} \begin{bmatrix} H_1 \\ \vdots \\ H_n \end{bmatrix} := \widetilde{F}^* T \widetilde{F}$$
$$S = \begin{bmatrix} S_1^* & \cdots & S_n^* \end{bmatrix} \begin{bmatrix} S_1 \\ \vdots \\ S_n \end{bmatrix} := G^* G.$$

Matrix sizes

- $\blacktriangleright H_k, S_k \in \mathbb{C}^{(2N_L) \times N_G}, T_k \in \mathbb{C}^{(2N_L) \times (2N_L)},$
- ► $F, G \in \mathbb{C}^{(2N_AN_L) \times N_G}$, $T \in \mathbb{C}^{(2N_AN_L) \times (2N_AN_L)}$.

Transform the problem!

Make T simpler

- ▶ the method can be applied even on already described matrices \widetilde{F} , G and T implicitely, but multiplication by T is slow
- T should be either factored (for example by using somewhat modified Hermitian indefinite factorization), or diagonalized (simultaneous diagonalization of T_ks) — diagonalization is too slow
- therefore, H is written as

$$H := F^*JF, \qquad J = \operatorname{diag}(\pm 1).$$

One-sided vs. two-sided method

- the original Hari–Zimmerman method works from both sides on the Hermitian matrix pair
- the modified method works from one side on the factors of the Hermitian matrix pair
- ▶ idea: think two-sided, act one-sided
- transformations will be computed from the pivot submatrices *H_{pq}* of *H* and *S_{pq}* of *S*

$$H_{pq} = \begin{bmatrix} F_p^* J F_p & F_p^* J F_q \\ & F_q^* J F_q \end{bmatrix}, \quad S_{pq} = \begin{bmatrix} G_p^* G_p & G_p^* G_q \\ & G_q^* G_q \end{bmatrix}.$$

The method consists of 3 transformations (Hari)

as a preprocessing step H and S can be scaled by the diagonal matrix D such that diag(DSD) = I

$$\begin{split} H_0 &:= DHD, \quad S_0 &:= DSD, \\ D &= \text{diag}\left(\frac{1}{\sqrt{s_{11}}}, \frac{1}{\sqrt{s_{22}}}, \dots, \frac{1}{\sqrt{s_{nn}}}\right) \end{split}$$

▶ in the first step the pivot submatrix \hat{S}_0 of S_0 is diagonalized by the complex rotation

$$\widehat{R}_1 = \begin{bmatrix} \cos \varphi_1 & e^{i\beta_1} \sin \varphi_1 \\ -e^{-i\beta_1} \sin \varphi_1 & \cos \varphi_1 \end{bmatrix},$$

The transformations

the first transformation is

$$H_1 = R_1^* H_0 R_1, \quad S_1 = R_1^* S_0 R_1,$$

 $R_1 = I$ except at the pivot positions, where $R_1 = \widehat{R}_1$.

• if *H* and *S* are preprocessed, then $\varphi_1 = -\frac{\pi}{4}$

- in the second step the diagonal of S_1 is rescaled to I
- this transformation is similar to the preprocessing step

 $H_2 := D_2 H_1 D_2, \quad S_2 := D_2 S_1 D_2.$

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The transformations

• in the third step the pivot submatrix \hat{H}_2 of H_2 is diagonalized by the complex rotation

$$\widehat{R}_3 = \begin{bmatrix} \cos\varphi_3 & e^{i\alpha_3}\sin\varphi_3 \\ -e^{-i\alpha_3}\sin\varphi_3 & \cos\varphi_3 \end{bmatrix},$$

the third transformation is

$$H_3 = R_3^* H_2 R_3, \quad S_3 = R_3^* S_2 R_3,$$

 $R_3 = I$ except at the pivot positions, where $R_3 = \hat{R}_3$. For if H and S are preprocessed, then $\varphi_3 = \vartheta + \frac{\pi}{4}$.

The transformations

• note that after the first three steps, the pivot submatrix \hat{S}_3 is still diagonal (in fact identity)

$$\widehat{S}_3 = \widehat{Z}^* \widehat{S} \widehat{Z}, \quad \widehat{Z} = \widehat{R}_1 \widehat{D}_2 \widehat{R}_3$$

if H and S are preprocessed, the fourth step is only formal it helps in coupling together all the transformations

$$H_4 = \Phi_4^* H_3 \Phi_4, \quad S_4 = \Phi_4^* S_3 \Phi_4, \quad \widehat{\Phi}_4 = \operatorname{diag}(e^{-i\sigma_p}, e^{-i\sigma_q}).$$

The coupled transformation $Z \dots$

▶ looks similar to an ordinary plane rotation: it is the identity matrix, except for its (p, q)-restriction Ẑ, where

$$\widehat{Z} = \frac{1}{\sqrt{1 - \left(|s_{pq}|\right)^2}} \begin{bmatrix} \cos\varphi & e^{i\alpha}\sin\varphi \\ -e^{-i\beta}\sin\psi & \cos\psi \end{bmatrix},$$

- φ and ψ are determined so that the transformations diagonalize the pivot submatrices \hat{H} and \hat{S}
- the transformation keeps the diagonal elements of S intact
- if S = I then Z is the ordinary rotation, the method is the ordinary Jacobi method for a single matrix.

Computation of the elements of \widehat{Z}

let

$$s = |s_{pq}|, \quad t = \sqrt{1 - s^2}, \quad r = s_{qq} - s_{pp},$$
$$\sigma = \begin{cases} 1 & e \ge 0\\ -1 & e < 0, \end{cases}, \quad u + iv = e^{-i \arg(s_{pq})} h_{pq},$$

▶ then if $\gamma = \alpha - \beta$

$$\tan(\gamma) = 2\frac{v}{r}, \qquad -\frac{\pi}{2} \le \gamma \le \frac{\pi}{2}$$
$$\tan(2\vartheta) = \sigma \frac{2u - (h_{pp} + h_{qq})s}{\sqrt{e^2 + 4v^2 \cdot t}}, \qquad -\frac{\pi}{4} < \vartheta \le \frac{\pi}{4}$$

Computation of the elements of \widehat{Z}

and

$$2\cos^2 \varphi = 1 + s\sin(2\vartheta) + t\cos(2\vartheta)\cos(\gamma), \quad 0 \le \varphi < \frac{\pi}{2}$$
$$2\cos^2 \psi = 1 - s\sin(2\vartheta) + t\cos(2\vartheta)\cos(\gamma), \quad 0 \le \psi < \frac{\pi}{2}$$
$$e^{i\alpha}\sin(\varphi) = \frac{(\sin(2\vartheta) - s) + i\sqrt{1 - s^2}\sin(\gamma)\cos(2\vartheta)}{1 - s\sin(2\vartheta) + \sqrt{1 - s^2}\cos(\gamma)\cos(2\vartheta)}$$
$$e^{-i\beta}\sin(\psi) = \frac{(\sin(2\vartheta) + s) - i\sqrt{1 - s^2}\sin(\gamma)\cos(2\vartheta)}{1 + s\sin(2\vartheta) + \sqrt{1 - s^2}\cos(\gamma)\cos(2\vartheta)}.$$

The pointwise algorithm

The implicit HZ algorithm Z = I: it = 0repeat // sweep loop it = it + 1for all pairs (p, q), 1compute $\widehat{H} = \begin{bmatrix} f_p^* J f_p & f_p^* J f_q \\ f_p^* J f_p \end{bmatrix}; \qquad \widehat{S} = \begin{bmatrix} g_p^* g_p & g_p^* g_q \\ g_p^* g_p & g_p^* g_q \end{bmatrix}$ compute the elements of \hat{Z} // transform F, G and Z $[f_p, f_a] = [f_p, f_a] \cdot \widehat{Z}$ $[g_p, g_q] = [g_p, g_q] \cdot \widehat{Z}$ $[z_n, z_n] = [z_n, z_n] \cdot \widehat{Z}$ until (no transf. in this sweep) or $(it \geq maxcyc)$

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Developer Edition of the Intel Xeon Phi 7210 (KNL) processor

- ▶ 96 GB of RAM per node,
- 64 cores per node,
- clock 1.30 GHz (Turbo Boost off),
- Intel AVX-512 (Advanced Vector Extensions) instruction set
- presence of two vector processing units (VPUs) per core each VPU operates independently on 512-bit vector registers – suitable for simultaneous processing of 16 single precision or 8 double precision numbers.

The first step

Hermitian indefinite factorization of all T_k 's

▶ for all T_k 's do in parallel

 $T_k = P_k^T R_k^* D_k R_k P_k,$

 P_k is a permutation matrix – formed as in LAPACK (as a sequence of partial permutations), D_k is block diagonal, with 1×1 or 2×2 diagonal blocks, R_k is upper triangular

• diagonalize all D_k 's in parallel

$$D_k = U_k^* \Delta_k U_k = U_k^* \sqrt{|\Delta_k| J_k \sqrt{|\Delta_k| U_k}},$$

 Δ_k diagonal, U_k block-diagonal, unitary, $J_k = \text{diag}(\pm 1)$,

The first step (cnt'd)

• for all J_k repermute them in parallel

$$J_k \mathrel{\mathop:}= \widetilde{P}_k^{\mathsf{T}} \operatorname{diag}(I, -I) \widetilde{P}_k$$

 \widetilde{P}_k is a permutation,

• multiply rows of all R_k and repermute them

$$R_k = \widetilde{P}_k \sqrt{|\Delta_k|} U_k$$

▶ repremute columns of all R_k according to pemutations stored in P_k

$$R_k := R_k P_k.$$

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Final state

$$T_k = R_k^* \operatorname{diag}(I, -I)R_k, \quad k = 1, \dots, n.$$

Comments

- ▶ in the first step, the factorization is sequential for each T_k
- each physical core of the Xeon Phi deals with one or more T_k in turn (OpenMP parallel do over all T_k)
- each core can use its own 1–4 hyperthreads in a call of the threaded BLAS routines – therefore even per core algoritm is somewhat parallel – but do not use hyperthreading, since...



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The second and the optional step

Form J, F and G

- store $J = diag(J_1, \ldots, J_n)$,
- multiply $F = diag(R_1, \ldots, R_n)F$, each R_k in parallel
- ▶ scale B_k by U_k in parallel and store G

Optional step — make J, F and G square

- square matrix faster HZ algorithm
- this step is the hyperbolic QR factorization on F and the QR factorization on G both algorithms moderately parallel
- pivoting strategy partial pivoting?, threshold pivoting?
- usage of (block)-reflectors or (block)-rotations?
- do it or not depends on the ratio (number of rows) / (number of columns) of F and G

Driver level of the implicit HZ algorithm

Details of the level-2 algorithm

- algorithm is Generalized Hyperbolic SVD of (F, G) with respect to J
- matrices F and G are divided in even number of block-columns

$$F = [F_1, \ldots, F_{2b}], \quad G = [G_1, \ldots, G_{2b}]$$

- number of block-columns depend on the number of physical cores of the processor (our case: 64 cores = maximum 128 blocks, no hyperthreading)
- each thread is connected to one physical core.

Driver level of the implicit HZ algorithm

Each thread . . .

- works on a pair of block-columns of each matrix given by some parallel pivot strategy
- allocates storage for [F_p, F_q], [G_p, G_q], their "shadow" counterparts, and for the part of the transformation matrix
- "shadow" memory used for scaling by J_k and data exchange
- since architecture is NUMA (Non Uniform Memory Access), columns are also physically copied to "shadow" memory (alternative: reassignment of pointers)
- allocates square space in fast MCDRAM for computation of the transformation Z_{pq} and the pivot block submatrices H_{pq} and S_{pq}.

Pivoting strategy

Parallel pivoting strategy

Choose pivot blocks independently in each step, for example, by using (block)-modulus strategy (not optimal!)





stopping criterion

- skip a transformation if cosines are 1
- final stop all transformations are skipped.

Driver level of the implicit HZ algorithm

Each thread . . .

- ▶ actually computes H_{pq} and S_{pq} (ZGEMM and ZHERK)
- factorizes H_{pq} and S_{pq} by the Hermitian indefinite factorization (test of definiteness of S_{pq})

$$H_{pq} = F_{pq}^* J_{pq} F_{pq}, \quad S_{pq} = G_{pq}^* G_{pq},$$

where F_{pq} , G_{pq} , and J_{pq} are square

- ▶ calls level-1 (non-blocked routine) on the triplet (F_{pq}, G_{pq}, J_{pq})
- applies transformation matrix to original F_{pq}, G_{pq}, and Z_{pq} (ZGEMMs)
- ▶ transfers one triplet of $(F_{\ell}, G_{\ell}, Z_{\ell})$, $\ell \in \{p, q\}$ to the next "owner" (thread) into its "shadow" memory.

Details of the level-1 algorithm

- single-threaded (including BLAS calls) SIMD-parallel code,
- the main loop sweep iterations (1, m, until convergence)
- parallel pivot strategy determines maximal number of independent pivot pairs — stage of the algorithm
- in each stage pairs are divided into groups of 8 pairs (AVX-512 instructions)
- compute 6 dot products (vectorized + reductions) with only 4 accesses of f_p, f_q, g_p, and g_q:

$$\widehat{H}_{pq} = \begin{bmatrix} f_p^* J_p f_p & f_p^* J_q f_q \\ & f_q^* J_q f_q \end{bmatrix}, \qquad \widehat{S}_{pq} = \begin{bmatrix} g_p^* g_p & g_p^* g_q \\ & g_q^* g_q \end{bmatrix},$$

Details of the level-1 algorithm

- an example: the dot products are computed without BLAS to avoid function calls (slow!)
- computing transformation matrices for 8 pairs simultaneously
- transformations to 8 column pairs (f_p, f_q), (g_p, g_q), (z_p, z_q) are applied sequentially for each pair (cache!)

Distribution of eigenvalues - NaCl



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Distribution of eigenvalues - AuAg



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Timings

			Number of cores	
Example	Problem	size	64	32
NaCl 2.5	50176 $ imes$	2256	800.70	556.95
NaCl 3.0	50176 $ imes$	3893	1973.64	1465.68
NaCl 3.5	50176 $ imes$	6217	2810.50	3660.44
NaCl 4.0	50176 $ imes$	9273	4846.98	7028.50
AuAg 2.5	$26136 \times$	3275	724.20	587.23
AuAg 3.0	$26136 \times$	5638	1549.92	1715.60
AuAg 3.5	$26136 \times$	8970	3152.78	4711.65
AuAg 4.0	26136 × 2	13379	6544.16	11955.74

Conclusion

On a particular hardware testing space is enormous

- use Quadrant or SNC-4 clustering mode?
- in a single step transform columns only once (block-oriented algorithm) or fully diagonalize them (full block algorithm)
- best pivoting strategy?
- is there need to shorten the columns by the hyperbolic QR factorization, and is there a switching point (use them or not)

Work in progress

. . .

- only lower 20% of the eigenvalues are needed
- is there any sufficiently parallel algorithm to compute them (without multiplication of the factors)?