

QMCkl: A Unified Approach to Accelerating Quantum Monte Carlo Codes

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The TREX European Center of Excellence





Partners

UNIVERSITY OF TWENTE.























Codes

- CHAMP
- QMC=Chem
- TurboRVB
- NECI
- QuantumPackage
- GammCor



CHAMP (Claudia Filippi)

- Wave function optimization: Jastrow, CI, MOs
- Ground/Excited states
- Geometry optimization



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- Molecular and Periodic systems
- JAGP, Pfaffian, ...
- LRDMC



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QMC=Chem (Michel Caffarel + Me!)

- DMC as "Post-Full-CI" energy calculations (CIPSI)
- Very large CI expansions (millions of determinants)
- Designed with HPC in mind
- Highly optimized with W. Jalby's group (UVSQ) in 2011-2013



- TREX CoE: Targeting REal chemical accuracy at the eXascale
- Started in Oct. 2020
- Objective: Make codes ready for exascale systems



- TREX CoE: Targeting REal chemical accuracy at the eXascale
- Started in Oct. 2020
- Objective: Make codes ready for exascale systems
- How: Instead of re-writing codes, provide libraries
 - One library for high-performance (QMCkl)
 - One library for exchanging information between codes (TREXIO)



The QMC kernel library (QMCkl)



- Progress in quantum chemistry requires codes with new ideas/algorithms
- New ideas/algorithms are implemented by physicists/chemists
- Different scientists have different programming language knowledge/preference
- Exascale machines are horribly complex to program



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Question

Is it reasonable to ask physicists/chemists to write codes for exascale machines?



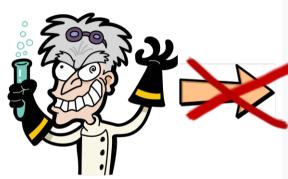
$$\mathsf{Z}_{n+1} = \mathsf{Z}_n + \mathsf{a}\mathsf{X}_n + \mathsf{Y}_n$$

```
do i=1,n
Z(i) = Z(i) + A * X(i) + Y(i)
end do
```

(from https://github.com/jeffhammond/dpcpp-tutorial)

```
std::vector<float> h X(length.xval):
std::vector<float> h Y(length, yval);
std::vector<float> h Z(length,zval);
trv {
    sycl::queue q(sycl::default selector{});
    const float A(aval):
    svcl::buffer<float.1> d X { h X.data(), svcl::range<1>(h X.size()) };
    svcl::buffer<float.1> d Y ( h Y.data(), svcl::range<1>(h Y.size()) }:
    sycl::buffer<float,1> d_Z { h_Z.data(), sycl::range<1>(h_Z.size()) };
   g.submit([&](svcl::handler& h) {
       auto X = d_X.template get_access<sycl::access::mode::read>(h);
       auto Y = d Y.template get access<svcl::access::mode::read>(h);
       auto Z = d Z.template get access<sycl::access::mode::read write>(h);
       h.parallel for<class nstream>( sycl::range<1>{length}, [=] (sycl::id<1> it) {
            const int i = it[0]:
           Z[i] += A * X[i] + Y[i]:
       });
      3):
     g.wait():
catch (sycl::exception & e) {
    std::cout << e.what() << std::endl;
   return 1:
```





```
std::vector<float> h_X(length,xval);
std::vector<float> h Y(length,vval);
std::vector<float> h Z(length,zval);
    sycl::queue q(sycl::default selector{});
   const float A(aval):
    sycl::buffer<float,1> d X { h X.data(), sycl::range<1>(h X.size()) };
   sycl::buffer<float,1> d_Y { h_Y.data(), sycl::range<1>(h_Y.size()) };
   sycl::buffer<float.1> d Z { h Z.data(), sycl::range<1>(h Z.size()) };
   q.submit([&](sycl::handler& h) {
       auto X = d_X.template get_access<sycl::access::mode::read>(h);
       auto Y = d Y.template get access<sycl::access::mode::read>(h);
       auto Z = d Z.template get access<svcl::access::mode::read write>(h):
       h.parallel_for<class nstream>( sycl::range<1>{length}, [=] (sycl::id<1> it) {
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       });
      });
     q.wait();
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    std::cout << e.what() << std::endl;
   return 1:
```

https://commons.wikimedia.org/wiki/File:Mad_scientist_transparent_background.svg

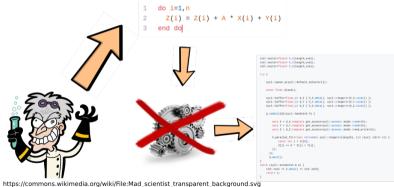


A compiler¹ that can read an average researcher's code and transform it into highly efficient code on an exascale machine.

¹Wikipedia: A compiler is a computer program that translates computer code written in one programming language (the source language) into another language (the target language)

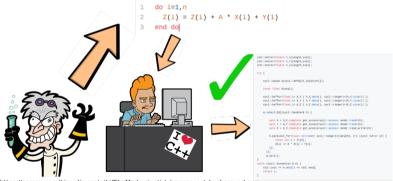


Artificial Intelligence was not ready in 2021 when we started the project . . .





... so we decided to use *Natural Intelligence*, and add a human layer between the machine and the researchers : a biological compiler



https://commons.wikimedia.org/wiki/File:Mad_scientist_transparent_background.svg



- Identify the common computational kernels of QMC
- Implement these kernels in a human-readable library (QMC experts)
- Bio-compile the human-readable library in a HPC-library (HPC experts)
- Scientists can link either library with their codes



For scientists

- The choice of the programming language is not imposed to the scientist
- The code can stay easy to understand by the physicists/chemists Performance-related aspects are delegated to the library
- Codes will not die with a change in hardware
- Scientific code development does not break the performance
- Scientists don't lose control on their codes



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Separation of concerns

- Scientists will never have to manipulate low-level HPC code
- HPC experts will not be required to be experts in theoretical physics
- Better re-use of the optimization effort among the community



The QMCkl Documentation library



- The API is C-compatible: QMCkl appears to scientists like a C library ⇒ can be used in all other languages
- System functions in programmed C (memory allocation, thread safety, etc)
- Computational kernels programmed in simple Fortran for readability
- A lot of documentation (remember: the HPC compiler is a human!)



Literate programming is a programming paradigm introduced by Donald Knuth in which a computer program is given an explanation of its logic in a natural language, such as English, interspersed with snippets of macros and traditional source code, from which compilable source code can be generated. (Wikipedia)



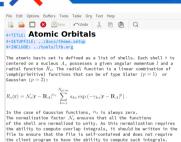
Literate programming with org-mode:

- Here, comments are more important than code
- Can add graphics, LATEXformulas, tables, etc
- Documentation always synchronized with the code
- Some functions can be generated by embedded scripts
- Web site auto-generated when code is pushed

Instead of writing comments documenting code, we write code illustrating documentation.



Literate programming with org-mode



Atomic orbitals (AOs) are defined as

 $y_i(\mathbf{r}) = P_{ero}(\mathbf{r}) R_{ero}(\mathbf{r})$

where $\theta(i)$ returns the shell on which the AO is expanded, and $\eta(i)$ denotes which angular function is chosen.

In this section we describe the kernels used to compute the values. gradients and Laplacian of the atomic basis functions.

```
@ Headers

⊗ Context...

@ Polynomial part...
Radial part
o Gaussian basis functions
   ~gmckl ao gaussian vgl~ computes the values, gradients and
   Laplacians at a given point of ~n~ Gaussian functions centered at
   the same point:
```

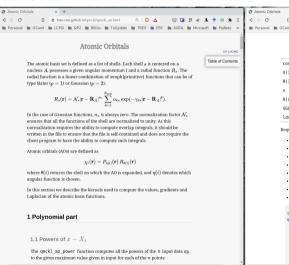
```
-context-
                     input | Global state
                     input | Array containing the coordinates of the points
       ~R(3)~
                    input | Array containing the x.v.z coordinates of the center
                    input | Number of computed Gaussians
                    input | Exponents of the Gaussians
       ~A(n)~
       ~VGI(ldv.5)~ | output | Value, gradients and Laplacian of the Gaussians
     ~ldv~ | input | Leading dimension of array ~VGL~
     Requirements:
    - -context- is not 8
    - -n- > 0
    - -1dv- >= 5
    - -A(i)- > 0 for all -i-
    - -X- is allocated with at least 3 x 8 bytes
    - -R- is allocated with at least 3 × 8 bytes
    - -A- is allocated with at least n \times 8 bytes
    - -VGL- is allocated with at least n \times 5 \times 8 bytes
#+begin src c :tangle (eval h func)
amckl exit code
qmckl_ao_gaussian_vgl(const qmckl_context context,
                     const double *X.
                     const double *R,
                     const int64 t *n.
                     const int64 t *A.
                     const double *VGL
                     const int64 t ldv):
    #+end src
    #+begin src f90 :tangle (eval f)
integer function qmckl_ao_gaussian_vgl_f(context, X, R, n, A, VGL, ldv) result(info)
 use omckl
 implicit none
 integer*8 , intent(in) :: context
 real*8 . intent(in) :: X(3). R(3)
  integer*8 . intent(in) :: n
 real*8 . intent(in) :: A(n)
 real*8 . intent(out) :: VGL(ldv.5)
 integer*8 , intent(in) :: ldv
 integer*8
                   0.14
 real+8
                   :: Y(3), r2, t, u, v
```





```
omckl an f.f98
                                                                              #define OMCKL INVALID CONTEXT
                                                                                                                      ((amckl exit code) 183)
                             amckl numbrec fh func.f90
gmckl ao fh func.f90
                             amckl numprec func.h
                                                                              #define OMCKL ALLOCATION FAILED
                                                                                                                      ((amckl exit code) 184)
qmckl_ao_func.h
                                                                              #define QMCKL_DEALLOCATION_FAILED
                                                                                                                      ((amckl exit code) 185)
gmckl_ao.org
                                                                              #define QMCKL_INVALID_EXIT_CODE
                                                                                                                      ((qmckl_exit_code) 186)
omckl ao private func.h
                             amckl numbrec type.h
                                                                              /* Context handling */
omckl an orivate type.h
                             amckl.org
amckl context.c
                             README, org
amckl context fh type.f90
                             test amckl
amckl context func.h
                             test amckl ac.c
amckl context.org
                             test muckl an f.f90
gmckl context private type.h test gmckl.c
amckl context type.h
                             test amckl context.c
amckl distance f.f90
                             test qmckl distance.c
amckl distance fh func.f90
                             test muckl distance f.f90
                                                                              /* #+NAME: gmckl context */
                             test_qmckl_error.c
                                                                              typedef int64 t amckl context :
gmckl distance.org
amckl error.c
                             test amckl numprec.c
                                                                              #define OMCKL NULL CONTEXT (amckl context) 0
amckl error fh func. f90
                             test amckl.org
(base) scenana@lngdb82:~/TRFX/gmckl/spc$
thank muckl get an basis shell and mom (const gmckl context context) {
 if (gmckl_context_check(context) == OMCKL_NULL_CONTEXT) {
                                                                                    #+NAME: MAX STRING LENGTH */
   return NULL:
                                                                              const char* qmckl string of error(const qmckl exit code error):
                                                                              void gmckl_string_of_error_f(const_gmckl_exit_code_error,
  if ( (ctx->ao basis.uninitialized & mask) != 0) {
   return NULL
 assert (ctx->ao_basis.shell_ang_mom != NULL);
                                                                                    explaining the error. The exit code can't be ~OMCKL SUCCESS~. */
~/TREX/gmckl/src/gmckl ag.c [unix] [C] [ 15% ] (104/674.16)
                                                                              ~/TREX/gmckl/include/gmckl.h [unix] [CPP] [ 31% ] (85/269.1)
                                            us * +6°C ♥ 1.1K - 5658 - LinksysRouter 82% A 49°C ♥ /: 216 1 188% ₩ 8 9 82% M 86/19 18:21 ₩ 10 80
```





□ 6 trex-coe.github.io/gmckl/gmckl_so... □ □ Δ □ B ← A + 9 9 5 = Be Personal Be O'Cami Be I CPO Be OP2 Be Riblio Be Tollodate Be TREX Be ERC Be AiDA Be Microsoft $\nabla_z v_i = -2a_i(X_z - R_z)v_i$ $\Delta v_i = a_i (4|X - R|^2 a_i - 6)v_i$ UPTHONE Table of Contents context input Global state X(3) input Array containing the coordinates of the points R(3) input Array containing the x.y.z coordinates of the center input Number of computed Gaussians A(n) input Exponents of the Gaussians VGL (1dv.5) output Value, gradients and Laplacian of the Gaussians input Leading dimension of array VGL Remirements: · context is not 0 n n n n ldv >= 5 A(i) > 0 for all i . X is allocated with at least 3 × 8 bytes . R is allocated with at least 3 × 8 bytes A is allocated with at least n × 8 bytes • VGL is allocated with at least $n \times 5 \times 8$ bytes gmckl_exit_code qmckl ao gaussian vgl(const qmckl context context, const double *X. const double *R. const int64 t *n. const int64 t *A. const double *VGL. const int64 t ldv):



At each QMC step, we need to evaluate $E_{loc}(\mathbf{r}_1,\ldots,\mathbf{r}_N) = \frac{\hat{H}\Psi(\mathbf{r}_1,\ldots,\mathbf{r}_N)}{\Psi(\mathbf{r}_1,\ldots,\mathbf{r}_N)}$:

- $\Psi(\mathsf{r}_1,\ldots,\mathsf{r}_N)$
- $\Delta_i \Psi(r_1, \ldots, r_i, \ldots, r_N)$: kinetic energy
- $\nabla_i \Psi(r_1, \ldots, r_i, \ldots, r_N)$: drift in the stochastic process



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Kernels implemented and well tested today

- AOs: $\chi(\mathbf{r}), \vec{\nabla}\chi(\mathbf{r}), \Delta\chi(\mathbf{r})$
- MOs: $\phi(\mathbf{r})$, $\vec{\nabla}\phi(\mathbf{r})$, $\Delta\phi(\mathbf{r})$
- Jastrow correlation factor (eN, ee, eeN)
- Inverses of small matrices



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Work in progress

■ Everything else required to compute Ψ , $\nabla \Psi$ and $\Delta \Psi$.

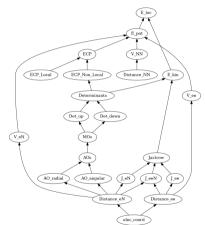


$$E_{loc}(R) = E_{pot}(R) + E_{kin}(R)$$

$$E_{pot}(R) = V_{ee}(R) + V_{eN}(R) + V_{NN}(R) + V_{ECP}(R)$$

$$E_{kin}(R) = -\frac{1}{2} \frac{\Delta \Psi(R)}{\Psi(R)}$$

$$\Psi(R) = \Phi(R)J(R)$$
...



All the graph is invalidated updated when the electron coordinates are changed.



Algorithms



Before computing anything, QMCkl needs to be given a trial wave function.

Setting wave function parameters

- Wave function exchange between codes is a major difficulty
- Our solution:
 - Define a standard format for wavefunction parameters
 - TREXIO: TREX Input/Output library (see Evgeny Posenitskiy's presentation)

Initialization of QMCkl

Two ways:

- 1 Control: Each array can be set by hand
- 2 Simplicity: Read all the wave function parameters from a TREXIO file



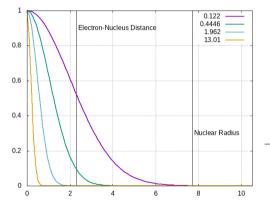
Atomic Orbitals

$$R_s(\mathbf{r}) = \mathcal{N}_s |\mathbf{r} - \mathsf{R}_A|^{n_s} \sum_{k=1}^{N_{\mathsf{prim}}} a_{ks} f_{ks} \exp\left(-\gamma_{ks} |\mathbf{r} - \mathsf{R}_A|^p\right).$$

Flexible

- Software like GAMESS use different normalization factors for *d* orbitals
- Implementing Slater-type orbitals is a minor modification (in the very long to-do list)
- Contribution from the FHI-AIMS group for the evaluation of numerical AOs
- Separation of the radial and angular components packed in shells
- Efficient computation of powers of x, y, z to maximize data re-use





- Definition of an atomic radius for each nucleus beyond which all AOs are zero (VGL^a).
- Primitives are sorted in ascending order of the exponents.
- Only non-zero elements are computed

^aVGL: value, gradients, Laplacian



Molecular Orbitals

$$\phi_{i}(\mathbf{r}_{j}) = \sum_{k} A_{ik} \chi_{k}(\mathbf{r}_{j}) \qquad B_{1} = A \cdot C_{1}$$

$$\nabla_{x} \phi_{i}(\mathbf{r}_{j}) = \sum_{k} A_{ik} \nabla_{x} \chi_{k}(\mathbf{r}_{j}) \qquad B_{2} = A \cdot C_{2}$$

$$\nabla_{y} \phi_{i}(\mathbf{r}_{j}) = \sum_{k} A_{ik} \nabla_{y} \chi_{k}(\mathbf{r}_{j}) \qquad B_{3} = A \cdot C_{3}$$

$$\nabla_{z} \phi_{i}(\mathbf{r}_{j}) = \sum_{k} A_{ik} \nabla_{z} \chi_{k}(\mathbf{r}_{j}) \qquad B_{4} = A \cdot C_{4}$$

$$\Delta \phi_{i}(\mathbf{r}_{j}) = \sum_{k} A_{ik} \Delta \chi_{k}(\mathbf{r}_{j}) \qquad B_{5} = A \cdot C_{5}$$





- QMC=Chem (2013): https://doi.org/10.1002/jcc.23216
- Exploits the common sparse character of the AO matrices:
 - When $\chi(r) = 0$ because r is too far, all the derivatives are also zero
 - Quadratic scaling
- Can be fully vectorized
 - >60% of peak performance on Sandy-Bridge CPUs

	Smallest system	β -Strand	β -Strand TZ	1ZE7	1AMB
N	158	434	434	1056	1731
$N_{ m basis}$	404	963	2934	2370	3892
% of non-zero ^a	81.3%	48.4%	73.4%	49.4%	37.1%
MO coefficients a_{ij}	(99.4%)	(76.0%)	(81.9%)	(72.0%)	(66.1%)
$(A_{ij} \neq 0)$					
Average % of non-zero					
basis functions $\chi_i(\mathbf{r}_j)$	36.2%	14.8%	8.2%	5.7%	3.9%
$(B_{1ij} \neq 0)$					
Average number of					
non-zero elements	146	142	241	135	152
per column of B_{1ij}					

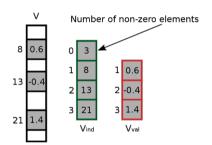


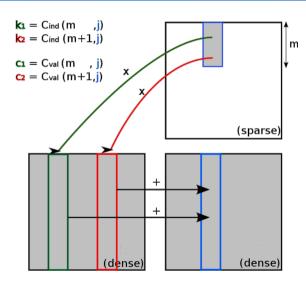


```
do j=1,point_num
1
         mo_vgl(:,:,j) = 0.d0
2
         do k=1.ao num
            if (ao_vgl(k,1,j) /= 0.d0) then
               c1 = ao_vgl(k,1,j)
               c2 = ao_vgl(k,2,j)
               c3 = ao_vgl(k,3,j)
               c4 = ao_vgl(k,4,j)
               c5 = ao_vgl(k,5,j)
               do i=1.mo num
10
                   mo_vgl(i,1,j) = mo_vgl(i,1,j) + coefficient_t(i,k) * c1
11
                   mo_vgl(i,2,j) = mo_vgl(i,2,j) + coefficient_t(i,k) * c2
12
                   mo_vgl(i,3,j) = mo_vgl(i,3,j) + coefficient_t(i,k) * c3
13
                   mo_vgl(i,4,j) = mo_vgl(i,4,j) + coefficient_t(i,k) * c4
14
                   mo_vgl(i,5,j) = mo_vgl(i,5,j) + coefficient_t(i,k) * c5
15
16
                end do
            end if
17
18
         end do
19
      end do
```



Sparse / dense matrix multiplication

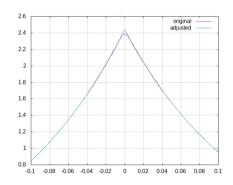






$$\phi_{\mathsf{cusp}\,i}(\mathsf{r}) = \phi_i(\mathsf{r}) - \phi_{\mathsf{s}_A}i(\mathsf{r}) + \sum_{l=0}^3 f_k \, |\mathsf{r} - \mathsf{R}_A|^k, \quad \mathsf{where} \, |\mathsf{r} - \mathsf{R}_A| < r_{\mathsf{cusp},A}$$

- $\phi_{s_A i}$: contributions of the s AOs centered at A to MO ϕ_i .
- 3 conditions:
 - Electron-nucleus cusp at $|r R_A| = 0$
 - Continuity of the MO: $\phi_{\text{cusp }i} = \phi_i$ when $|\mathbf{r} \mathbf{R}_A| = r_{\text{cusp }A}$
 - Continuity of the gradient: $\nabla \phi_{\text{cusp }i}(\mathbf{r}) = \nabla \phi_i(\mathbf{r})$ when $|\mathbf{r} \mathsf{R}_A| = r_{\text{cusp},A}$





$$J_{\text{een}}(\mathsf{r},\mathsf{R}) = \sum_{\alpha=1}^{N_{\text{nucl}}} \sum_{i=1}^{N_{\text{elec}}} \sum_{j=1}^{i-1} \sum_{p=2}^{N_{\text{nord}}} \sum_{k=0}^{p-1} \sum_{l=0}^{p-k-2\delta_{k,0}} c_{lkp\alpha} \left(r_{ij} \right)^k \left[\left(R_{i\alpha} \right)^l + \left(R_{j\alpha} \right)^l \right] \left(R_{i\alpha} R_{j\alpha} \right)^{(p-k-l)/2}$$

can be rewritten as

$$J_{\text{een}}(\mathbf{r},\mathsf{R}) = \sum_{p=2}^{N_{\text{nord}}} \sum_{k=0}^{p-1} \sum_{l=0}^{p-k-2\delta_{k,0}} \sum_{\alpha=1}^{N_{\text{nucl}}} c_{lkp\alpha} \sum_{i=1}^{N_{\text{elec}}} \bar{\mathbf{R}}_{i,\alpha,(p-k-l)/2} \, \bar{\mathbf{P}}_{i,k,\alpha,(p-k+l)/2} \, (\downarrow \text{ complexity})$$

with

$$ar{\mathtt{P}}_{i,k,lpha,I} = \sum_{i=1}^{N_{\mathsf{elec}}} ar{\mathtt{r}}_{i,k,j} \; ar{\mathtt{R}}_{j,lpha,I}. \; \mathsf{(GEMM)}$$

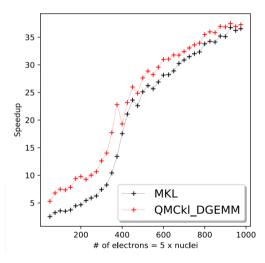


$$\nabla_{im} J_{\text{een}}(\mathbf{r}, \mathsf{R}) \ = \ \sum_{p=2}^{N_{\text{nord}}} \sum_{k=0}^{p-1} \sum_{l=0}^{p-k-2\delta_{k,0}} \sum_{\alpha=1}^{N_{\text{nucl}}} c_{lkp\alpha} \sum_{i=1}^{N_{\text{elec}}} \bar{\mathbf{G}}_{i,m,\alpha,(p-k-l)/2} \bar{\mathbf{P}}_{i,\alpha,k,(p-k-l)/2} + \\ \bar{\mathbf{G}}_{i,m,\alpha,(p-k+l)/2} \bar{\mathbf{P}}_{i,\alpha,k,(p-k-l)/2} + \bar{\mathbf{R}}_{i,\alpha,(p-k-l)/2} \bar{\mathbf{Q}}_{i,m,\alpha,k,(p-k-l)/2} + \\ \bar{\mathbf{R}}_{i,\alpha,(p-k+l)/2} \bar{\mathbf{Q}}_{i,m,\alpha,k,(p-k-l)/2} + \delta_{m,4} (\\ \bar{\mathbf{G}}_{i,1,\alpha,(p-k+l)/2} \bar{\mathbf{Q}}_{i,1,\alpha,k,(p-k-l)/2} + \bar{\mathbf{G}}_{i,2,\alpha,(p-k+l)/2} \bar{\mathbf{Q}}_{i,2,\alpha,k,(p-k-l)/2} + \\ \bar{\mathbf{G}}_{i,3,\alpha,(p-k+l)/2} \bar{\mathbf{Q}}_{i,3,\alpha,k,(p-k-l)/2} + \bar{\mathbf{G}}_{i,1,\alpha,(p-k-l)/2} \bar{\mathbf{Q}}_{i,1,\alpha,k,(p-k+l)/2} + \\ \bar{\mathbf{G}}_{i,2,\alpha,(p-k-l)/2} \bar{\mathbf{Q}}_{i,2,\alpha,k,(p-k-l)/2} + \bar{\mathbf{G}}_{i,3,\alpha,(p-k-l)/2} \bar{\mathbf{Q}}_{i,3,\alpha,k,(p-k+l)/2})$$

with

$$\bar{\mathbf{g}}_{i,m,\alpha,l} = \frac{\partial \left(R_{i\alpha}\right)^l}{\partial r_i}, \qquad \bar{\mathbf{g}}_{i,m,j,k} = \frac{\partial \left(r_{ij}\right)^k}{\partial r_i}, \qquad \text{ and } \bar{\mathbf{Q}}_{i,m,\alpha,k,l} = \sum_{i=1}^{N_{\mathsf{elec}}} \bar{\mathbf{g}}_{i,m,j,k} \; \bar{\mathbf{R}}_{j,\alpha,l}$$







HPC implementations



- MAQAO, developed by the UVSQ team, is used to help us optimize the CPU code
 - Loop-level diagnostics
 - Vectorization ratio
 - Hints to improve efficiency
- Algorithms rewritten in C:
 - C compilers are usually more mature than Fortran on new hardware
 - Access to more low-level features than Fortran (pinned memory, alignment, inline assembly, etc)
- Precision can be changed on-the-fly: switch to single-precision if possible
- Specialization:
 - Specialization for s, p and d AOs
 - Inverse of small matrices hard-coded for 2×2 to 5×5
 - Small matrix multiplication
 -



```
subroutine cofactor4(a.LDA.b.LDB.na.det 1)
   Amplicit none
   double precision, intent(in) :: A (LDA,na)
   double precision, intent(out) :: B (LDA,na)
   integer*8, intent(in)
                                                                  :: LDA. LDB
   integer*8, intent(in)
                                                                  :: na
   double precision, intent(inout) :: det_l
   integer :: i.i
   det_l = a(1,1)*(a(2,2)*(a(3,3)*a(4,4)-a(3,4)*a(4,3)) &
                                    -a(2,3)*(a(3,2)*a(4,4)-a(3,4)*a(4,2)) &
                                    +a(2.4)*(a(3.2)*a(4.3)-a(3.3)*a(4.2))) &
                   -a(1,2)*(a(2,1)*(a(3,3)*a(4,4)-a(3,4)*a(4,3)) &
                                    -a(2,3)*(a(3,1)*a(4,4)-a(3,4)*a(4,1)) &
                                    +a(2.4)*(a(3.1)*a(4.3)-a(3.3)*a(4.1))) &
                   +a(1,3)*(a(2,1)*(a(3,2)*a(4,4)-a(3,4)*a(4,2)) &
                                    -a(2,2)*(a(3,1)*a(4,4)-a(3,4)*a(4,1)) &
                                    +a(2,4)*(a(3,1)*a(4,2)-a(3,2)*a(4,1))) &
                   -a(1,4)*(a(2,1)*(a(3,2)*a(4,3)-a(3,3)*a(4,2)) &
                                    -a(2,2)*(a(3,1)*a(4,3)-a(3,3)*a(4,1)) &
                                    +a(2,3)*(a(3,1)*a(4,2)-a(3,2)*a(4,1)))
   b(1,1) = a(2,2) * (a(3,3) * a(4,4) - a(3,4) * a(4,3)) - a(2,3) * (a(3,2) * a(4,4) - a(3,4) * a(4,2)) + a(2,4) * (a(3,2) * a(4,3) - a(3,3) * a(4,2))
   b(2,1) = -a(2,1)*(a(3,3)*a(4,4)-a(3,4)*a(4,3))*a(2,3)*(a(3,1)*a(4,4)-a(3,4)*a(4,1))-a(2,4)*(a(3,1)*a(4,3)-a(3,3)*a(4,1))
   b(3,1) = a(2,1)*(a(3,2)*a(4,4)-a(3,4)*a(4,2))-a(2,2)*(a(3,1)*a(4,4)-a(3,4)*a(4,1))+a(2,4)*(a(3,1)*a(4,2)-a(3,2)*a(4,1))
   b(4,1) = -a(2,1) \\ \times (a(3,2) \\ \times a(4,3) \\ -a(3,3) \\ \times a(4,2) \\ +a(2,2) \\ \times (a(3,1) \\ \times a(4,3) \\ -a(3,3) \\ \times a(4,1) \\ -a(2,3) \\ \times (a(3,1) \\ \times a(4,2) \\ -a(3,2) \\ \times a(4,1) \\ -a(2,3) \\ \times (a(3,1) \\ \times a(4,2) \\ -a(3,2) \\ \times a(4,3) \\ -a(3,3) \\ \times a(4,3) \\ -a(4,3) \\ 
   b(1,2) = -a(1,2)*(a(3,3)*a(4,4)-a(3,4)*a(4,3))+a(1,3)*(a(3,2)*a(4,4)-a(3,4)*a(4,2))-a(1,4)*(a(3,2)*a(4,3)-a(3,3)*a(4,2))
   b(2,2) = a(1,1)*(a(3,3)*a(4,4)-a(3,4)*a(4,3))-a(1,3)*(a(3,1)*a(4,4)-a(3,4)*a(4,1))+a(1,4)*(a(3,1)*a(4,3)-a(3,3)*a(4,1))
   b(3,2) = -a(1,1) + (a(3,2) + a(4,4) - a(3,4) + a(4,2) + a(1,2) + (a(3,1) + a(4,4) - a(3,4) + a(4,1)) - a(1,4) + (a(3,1) + a(4,2) - a(3,2) + a(4,1))
   b(4,2) = a(1,1) * (a(3,2) * a(4,3) - a(3,3) * a(4,2)) - a(1,2) * (a(3,1) * a(4,3) - a(3,3) * a(4,1)) + a(1,3) * (a(3,1) * a(4,2) - a(3,2) * a(4,1))
   b(1,3) = a(1,2) * (a(2,3) * a(4,4) - a(2,4) * a(4,3)) - a(1,3) * (a(2,2) * a(4,4) - a(2,4) * a(4,2)) + a(1,4) * (a(2,2) * a(4,3) - a(2,3) * a(4,2))
   b(2,3) = -a(1,1)*(a(2,3)*a(4,4)-a(2,4)*a(4,3))+a(1,3)*(a(2,1)*a(4,4)-a(2,4)*a(4,1))-a(1,4)*(a(2,1)*a(4,3)-a(2,3)*a(4,1))
   b(3,3) = a(1,1) * (a(2,2) * a(4,4) - a(2,4) * a(4,2) - a(1,2) * (a(2,1) * a(4,4) - a(2,4) * a(4,1)) + a(1,4) * (a(2,1) * a(4,2) - a(2,2) * a(4,1))
   b(4,3) = -a(1,1) * (a(2,2) * a(4,3) - a(2,3) * a(4,2)) + a(1,2) * (a(2,1) * a(4,3) - a(2,3) * a(4,1)) - a(1,3) * (a(2,1) * a(4,2) - a(2,2) * a(4,1))
   b(1,4) = -a(1,2) \times (a(2,3) \times a(3,4) - a(2,4) \times a(3,3)) + a(1,3) \times (a(2,2) \times a(3,4) - a(2,4) \times a(3,2)) - a(1,4) \times (a(2,2) \times a(3,3) - a(2,3) \times a(3,2))
   b(2,4) = a(1,1) * (a(2,3) * a(3,4) - a(2,4) * a(3,3)) - a(1,3) * (a(2,1) * a(3,4) - a(2,4) * a(3,1)) + a(1,4) * (a(2,1) * a(3,3) - a(2,3) * a(3,1))
   b(3,4) = -a(1,1) \times (a(2,2) \times a(3,4) - a(2,4) \times a(3,2)) + a(1,2) \times (a(2,1) \times a(3,4) - a(2,4) \times a(3,1)) - a(1,4) \times (a(2,1) \times a(3,2) - a(2,2) \times a(3,1))
   b(4,4) = a(1,1) * (a(2,2) * a(3,3) - a(2,3) * a(3,2)) - a(1,2) * (a(2,1) * a(3,3) - a(2,3) * a(3,1)) + a(1,3) * (a(2,1) * a(3,2) - a(2,2) * a(3,1))
end subroutine cofactor4
```



- GPU library has the same functions, suffixed with _device
- Two different flavours: OpenMP or OpenACC
- Possibility to use CPU and GPU library together in the same code
- In early development, not fully integrated to our codes yet (work in progress)
- Although the kernels are fast on Nvidia GPUs, GPU acceleration is not clear because of data transfer
 - Maybe efficient on next generation of hardware
- On GPU, brute-force CuBLAS DGEMM is faster than sparse AO-MO transformation. Energy efficiency?



- Tensor core instructions are not generated in OpenMP kernels ⇒≤ 50% peak DP
- Conflict between OpenMP runtime of the code and of QMCkl-GPU ⇒
 - Need to compile the code with GPU compiler (Nvfortran)
 - May not compile, or with low CPU efficiency
 - Our solution: decouple QMCkl-CPU and QMCkl-GPU and recover CPU performance with QMCkl-CPU
- $lue{}$ RocBLAS \sim CuBLAS, but some OpenMP kernels have 10 imes lower performance on AMD GPUs than Nvidia (under investigation. . .)
- Unreliable software stack: ⇒ Compared to CPU, very inefficient in human resources
- Open Question:
 - Should we have opted instead for vendor-specific implementations? (Cuda, HIP)



```
1  $ tar -zxvf qmckl.tar.gz
2  $ cd qmckl
3  $ ./configure --enable-hpc
4  $ make -j 32
5  $ make check
6  $ make install
```

- QMCkl has been used in
 - C / C++
 - Fortran
 - Python
 - Julia
 - Rust

- Very few dependencies:
 - BLAS/Lapack (CPU)
 - TREXIO (optional) with HDF5 (optional)
- BSD license: very permissive. You can distribute the tar.gz with your code
- Hosted on GitHub: https://github.com/trex-coe/qmckl



Integration into TREX codes



- Single-core benchmark: C₆₀, Hartree-Fock/cc-pVQZ/ECP(BFD)
 - Time for a single MC step (all-electrons)
 - 4140 AOs, 120 MOs, 240 electrons

CPU	Compiler	QMCkl	milliseconds	Speedup
Intel(R) Core(TM) i7	ifort/mkl	-	24.58	
(8-core Laptop, 2.8GHz)	ifort/mkl	gcc12	24.06	1.02x
	ifort/mkl	icx	23.85	1.03x
	gfortran/openblas	-	30.58	
	gfortran/openblas	gcc12	26.04	1.17×
ARM Neoverse V1	gfortran/armpl	-	41.24	
(80 cores, 3GHz)	gfortran/armpl	gcc12	31.91	1.29x



- Single-core benchmark: C₆₀, Hartree-Fock/cc-pVXZ/ECP(BFD)
 - Short VMC run
 - 4140 AOs, 120 MOs, 240 electrons

Basis	# AOs	Compiler	QMCkl	seconds	Speedup
cc-pVDZ	840	ifort/mkl	-	315.45	
			gcc12	218.29	1.45x
			icx	212.35	1.49×
cc-pVTZ	2040	ifort/mkl	-	565.67	
			gcc12	287.32	$1.97 \times$
			icx	271.68	2.08x
cc-pVQZ	4140	ifort/mkl	-	993.42	
			gcc12	462.74	2.15x
			icx	441.32	2.25x



Other possible applications beyond accelerating QMC

- Reproducibility of QMC calculations (Jastrow factors)
- 3D visualization software:
 - AO or MO visualization
 - Interpretative methods like AIM or ELF
- Numerical integration
 - Computation of density grids for DFT with gradients
 - Jastrow factor in transcorrelated methods (Quantum Package)
- Teaching QMC algorithms in Jupyter notebooks
- Implementation of QMC methods in traditional quantum chemistry software



Example: Evaluate MOs on a grid

```
import qmckl
    import numpy as np
3
    def main(trexio filename):
      context = gmckl.context_create()
                                                          # Create a QMCkl context
5
      qmckl.trexio_read(context, trexio_filename)
                                                          # Read the TREXIO file into the context
6
      nucl_num = amckl.get_nucleus_num(context)
                                                                     # Get the number of nuclei
9
      nucl_coord = qmckl.get_nucleus_coord(context, 'N', nucl_num*3) # Get the nuclear coordinates
      nucl_coord = np.reshape(nucl_coord, (3, nucl_num))
10
      mo\_num
                 = gmckl.get_mo_basis_mo_num(context)
                                                                     # Get the number of MOs
11
12
13
      point
                = setup_grid_points(nucl_coord)
      point_num = len(point)
14
15
      qmckl.set_point(context, 'N', point_num, np.reshape(point, (point_num*3))) # Give points to QMCkl
16
17
      mo_value = qmckl.get_mo_basis_mo_value(context, point_num*mo_num) # Get the values of the MOs
18
19
      gmckl.context_destroy(context)
20
                                                       # Free QMCkl resources
```



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