Software Projects: Examples from Nuclear Physics

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Background: Infinite nuclear matter

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Case I: Code optimization and parallelization

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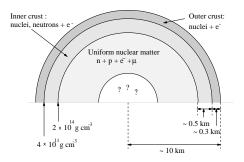
Case II: Towards object-oriented Fortran code

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We want to model a neutron star!



- ▶ A massive star burns up its fuel \rightarrow Supernova explosion \rightarrow A neutron star is formed
- Microscopic nuclear matter calculations give input to the neutron star Equation of State (EOS)

In mathematical terms...

We need to solve the eigenvalue problem

$$\hat{H}\Psi = E\Psi$$
,

where E is the total energy and the Hamiltonian operator

$$\hat{H} = \hat{T} + \hat{V}$$

$$= -\frac{1}{2m} \sum_{i=1}^{A} \nabla_i^2 + \sum_{i < i}^{A} \hat{v}(\mathbf{r}_i, \mathbf{r}_j),$$

 $\hat{T}=$ kinetic energy operator, $\hat{V}=$ two-particle interaction operator,

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What needs to be coded

Expression for the correlation energy: $(\mathbf{k} = (k_x, k_y, k_z))$

$$\underline{\Delta E_{Correlation}} = \int d\mathbf{k}_{i} \int d\mathbf{k}_{j} \int d\mathbf{k}_{a} \int d\mathbf{k}_{b}$$
The eigenvalue
$$\times \underbrace{\langle \mathbf{k}_{i} \mathbf{k}_{j} | v | \mathbf{k}_{a} \mathbf{k}_{b} \rangle}_{\text{Known, not closed form}} \underbrace{\langle \mathbf{k}_{a} \mathbf{k}_{b} | t | \mathbf{k}_{i} \mathbf{k}_{j} \rangle}_{\text{The unknown: A dense, block-diagonal gigantic matrix}} (1)$$

$$0 = \langle \mathbf{k}_{a} \mathbf{k}_{b} | v | \mathbf{k}_{i} \mathbf{k}_{j} \rangle + \Delta \varepsilon (\mathbf{k}_{i}, \mathbf{k}_{j}, \mathbf{k}_{a}, \mathbf{k}_{b}) \langle \mathbf{k}_{a} \mathbf{k}_{b} | t | \mathbf{k}_{i} \mathbf{k}_{j} \rangle + \int d\mathbf{k}_{c} \int \mathbf{k}_{d} \langle \mathbf{k}_{a} \mathbf{k}_{b} | v | \mathbf{k}_{c} \mathbf{k}_{d} \rangle \langle \mathbf{k}_{c} \mathbf{k}_{d} | t | \mathbf{k}_{i} \mathbf{k}_{j} \rangle,$$
(2)

- We use the Picard iteration method
- ► Conversion to quantum number basis (Fourier grid):

$$\mathbf{k} \rightarrow \{k(IS)\mathcal{J}m_{\mathcal{J}}M_{\mathcal{T}}\}$$



Challenges

The most naive implementation:

Requires far too much RAM memory and CPU time

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- Faster matrix operations with MKL BLAS than with standard BLAS
- Utilize symmetries in physics (e.g. in total momentum and different quantum numbers):
 - ⇒ Matrices become block diagonal
- ▶ I stored non-zero block diagonals into matrix lists

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- Don't calculate the same thing many times:
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- ▶ If tests consume much time, use sparingly

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Parallelization

- ► The matrix lists were divided to different MPI processes
 - \Rightarrow available memory from many computing nodes
 - ⇒ many processors available
- Most of the CPU time was used to set up a big matrix list before the Picard iterations
 - The matrix list setup was parallelized on each node using OpenMP

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- ► The previous code was written in Fortran 90/95 using modules
- ► The Fortran 2003 standard fully supports object-oriented programming
- ► I have written a code for one-dimensional nuclear matter in (almost) object-oriented Fortran

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- ► The code utilizes inheritance of an abstract class, as well as dynamic polymorphism (virtual methods)

Example of an abstract class

```
TYPE, ABSTRACT :: CorrEnergy
    TYPE(Hamiltonian) :: hamilt
    TYPE(SpPotential) :: spPot
    INTEGER :: nkh.nkp.nkCM
    TYPE(QuadList) :: quads
    REAL(DP) :: eneCorr
  CONTAINS
    PROCEDURE(calcCorrEne). &
          DEFERRED :: calcCorrEnergy
 END TYPE CorrEnergy
 ABSTRACT INTERFACE
    SUBROUTINE calcCorrEne(eneCorr)
      IMPORT CorrEnergy
      CLASS(CorrEnergy), TARGET, INTENT(INOUT) :: eneCorr
    END SUBROUTINE calcCorrEne
 END INTERFACE
CONTAINS
 SUBROUTINE printCorrEnergy(eneRef,eneCorrel)
   TYPE(RefEnergy), INTENT(IN) :: eneRef
   CLASS(CorrEnergy), POINTER, INTENT(IN) :: eneCorrel
   REAL(DP) :: eneTot
```

A routine that uses polymorphism

```
SUBROUTINE calcPt2Energy(eneCorr)
 USE OuadratureMod
 CLASS(Pt2CorrEnergy), TARGET, INTENT(INOUT) :: eneCorr
 TYPE(Quadrature) :: relmomH, relmomP1, relmomP2, CMmom
 REAL(DP), ALLOCATABLE, DIMENSION(:) :: kpPoints, weightKp
 REAL(DP) :: PMax,interaction,kh,wkh,kp,wkp,P,wP
 REAL(DP) :: spPotK1,spPotK2,spPotK3,spPotK4
 REAL(DP) :: temp.khMax.kpMin.kl.k2.k3.k4
  INTEGER :: iP.ikh.ikp
 ALLOCATE(kpPoints(2*eneCorr%nkp),&
      weightKp(2*eneCorr%nkp))
 kpPoints = 0.d0
 weightKp = 0.d0
 PMax = 2.d0*eneCorr%hamilt%kF
 CMmom = gaussLegendre(-Pmax,Pmax,eneCorr%nkCM)
 temp = 0.d0
 DO iP=1, eneCorr%nkCM
    P = CMmom%xpoints(iP)
    wP = CMmom%weightx(iP)
```

Code structure when using objects

```
SUBROUTINE setupSpPotMesh(spPot)
  TYPE(SpPotential), INTENT(INOUT) :: spPot
  TYPE(Quadrature) :: labmom
  REAL(DP) :: kmax,k,potential
  INTEGER :: ik
  kmax = spPot%hamilt%krelCutoff + spPot%hamilt%kF
  labmom = gaussLegendre(-kmax,kmax,spPot%mesh%nMeshp)
  spPot%mesh%points = labmom%xpoints
  DO ik=1, spPot%mesh%nMeshp
     k = spPot%mesh%points(ik)
     potential = calcSpPotential(k,spPot)
     spPot%mesh%values(ik) = potential
  ENDDO
END SUBROUTINE setupSpPotMesh
```

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