Solving Many-Body Schrödinger Equation Using Density Functional Theory and Finite Elements

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Introduction

Contens

- Density Functional Theory
- Spherically symmetric problems, examples
- Mixing schemes
- Non symmetric 3D problems, Finite Element Method

Schrödinger equation

$$\hat{H}\ket{\Psi} = (\hat{T} + \hat{U} + \hat{V})\ket{\Psi} = E\ket{\Psi}$$

where

$$\hat{T} = \sum_{i=1}^{N} -\frac{1}{2} \nabla_{i}^{2}$$

$$\hat{U} = \sum_{i < j} U(\mathbf{r_i}, \mathbf{r_j}) = \frac{1}{2} \sum_{i,j} U(\mathbf{r_i}, \mathbf{r_j})$$

$$U(\mathbf{r_i}, \mathbf{r_j}) = U(\mathbf{r_j}, \mathbf{r_i}) = \frac{1}{|\mathbf{r_i} - \mathbf{r_j}|}$$

$$\hat{V} = \sum_{i=1}^{N} v(\mathbf{r_i})$$

$$v(\mathbf{r_i}) = \sum_{k=1}^{N} -\frac{Z_k}{|\mathbf{r_i} - \mathbf{R_k}|}$$

Density Functional Theory

We solve the Kohn-Sham equations:

$$(-\frac{1}{2}\nabla^2 + V_H(\mathbf{r}) + V_{xc}(\mathbf{r}) + v(\mathbf{r}))\psi_i(\mathbf{r}) = \epsilon_i \psi(\mathbf{r})$$

that yield the orbitals ψ_i that reproduce the density $n(\mathbf{r})$ of the original interacting system

$$n(\mathbf{r}) = \sum_{i}^{N} |\psi_{i}(\mathbf{r})|^{2}$$

$$V_{H}(\mathbf{r}) = \frac{\delta E_{H}}{\delta n(\mathbf{r})} = \frac{1}{2} \int \frac{n(\mathbf{r}')}{|\mathbf{r} - \mathbf{r}'|} d^{3}r' \iff \nabla^{2}V_{H} = n(\mathbf{r})$$

$$E_{xc}[n] = (T + U)[n] - E_{H}[n] - T_{S}[n]$$

$$V_{xc}(\mathbf{r}) = \frac{\delta E_{xc}[n]}{\delta n(\mathbf{r})}$$

$$v(\mathbf{r}) = \sum_{k} -\frac{Z_{k}}{|\mathbf{r} - \mathbf{R}_{k}|}$$

Atomic wavefunctions

Spherically symmetric potential:

$$V(\mathbf{x}) = V(r)$$

$$\psi_{nlm}(\mathbf{x}) = R_{nl}(r) Y_{lm}\left(\frac{\mathbf{x}}{r}\right)$$

Radial Schrödinger equation:

$$R''_{nl} + \frac{2}{r}R'_{nl} + \frac{2M}{\hbar^2}(E - V)R_{nl} - \frac{I(I+1)}{r^2}R_{nl} = 0$$

Relativistic atomic wavefunctions

Dirac equation:

$$(ic\gamma^{\mu}D_{\mu}-mc^{2})\psi=0$$
 $D_{\mu}=\partial_{\mu}+ieA_{\mu}$

Radial Dirac equation:

$$g_{\kappa}^{"} + \left(\frac{2}{r} + \frac{V^{\prime}}{2Mc^{2}}\right)g_{\kappa}^{\prime} + \left[\left(E - V\right) - \frac{\kappa(\kappa + 1)}{2Mr^{2}} + \frac{\kappa + 1}{4M^{2}c^{2}r}V^{\prime}\right]2Mg_{\kappa} = 0$$

$$f_{\kappa} = \frac{g_{\kappa}^{\prime}}{2Mc} + \frac{\kappa + 1}{r}\frac{g_{\kappa}}{2Mc}$$

$$R^{2} = f^{2} + \sigma^{2}$$

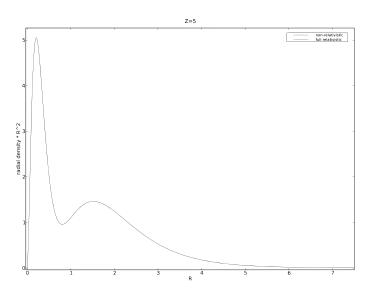
Radial Schrödinger equation:

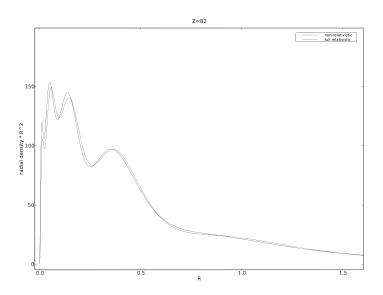
$$R'' + \frac{2}{r}R' + \left[(E - V) - \frac{I(I+1)}{2Mr^2} \right] 2MR = 0$$



Code (300 lines in Python, 800 in Fortran)

```
from atom import atom, show
import radial
import utils
def do(Z):
    R = radial.create_log_grid(Z)
    s = atom(Z,alpha=0.3,iter=20,relat=0,grid=R)
    n5_lda = radial.KS_construct_density(s,R,Z)*R*R
    s = atom(Z,alpha=0.3,iter=20,relat=2,grid=R)
    n5_rlda = radial.KS_construct_density(s,R,Z)*R*R
    utils.makeplot(R,[
        (n5_lda, "b-", "non-relativistic"),
        (n5_rlda, "g-", "full relativistic"),
        ],title="Z=%d"%(Z),xleg="R",
        yleg="radial density * R^2")
do(5)
#do(82)
```





Lead - nonrelativistic calculation

```
2s(2): -488.8433352
                           2p(6): -470.8777849
                           3s(2): -116.526852
                           3p(6): -107.950391
Iterations: 20
                           3d(10): -91.88992429
|F(x)| = 0.00003516
                           4s(2): -25.75333021
Agrees with NIST:
                           4p(6): -21.99056413
http://physics.nist.gov/
                           4d(10): -15.03002657
                           4f(14): -5.592531664
                           5s(2): -4.206797624
                           5p(6): -2.941656967
                           5d(10): -0.9023926829
                           6s(2): -0.3571868295
                           6p(2): -0.1418313263
```

1s(2): -2901.078061

Lead - relativistic calculation

```
4s(2) j=1+1/2: -31.15015728
Iterations: 20
                             4p(6) j=1-1/2: -26.73281564
|F(x)| = 0.00000584
                             4p(6) j=1+1/2: -22.38230707
                             4d(10) j=1-1/2: -15.1647618
1s(2) j=1+1/2: -3209.51946
                             4d(10) j=1+1/2: -14.3484973
2s(2) j=1+1/2: -574.1825655
                             5s(2) j=1+1/2: -5.225938506
2p(6) j=1-1/2: -551.7234408
                             4f(14) j=1-1/2: -4.960490099
2p(6) j=1+1/2: -472.3716103
                             4f(14) j=1+1/2: -4.775660273
3s(2) j=1+1/2: -137.8642241
                             5p(6) j=1-1/2: -3.710458943
3p(6) j=1-1/2: -127.6789451
                             5p(6) j=1+1/2: -2.889127431
3p(6) j=1+1/2: -109.9540395
                             5d(10) j=1-1/2: -0.8020049565
3d(10) j=1-1/2: -93.15817605
                             5d(10) j=1+1/2: -0.7070299184
3d(10) j=1+1/2: -89.36399096
                             6s(2) j=1+1/2: -0.4209603386
                             6p(2) j=1-1/2: -0.1549640727
```

Iteration to self-consistency

The problem:

$$F(x) = x$$

equivalently

$$\mathbf{R}(\mathbf{x}) = 0$$

for $\mathbf{R}(\mathbf{x}) = \mathbf{F}(\mathbf{x}) - \mathbf{x}$. We approximate

$$\mathsf{R}(\mathsf{x}_{M+1}) - \mathsf{R}(\mathsf{x}_M) pprox \mathsf{J} \cdot (\mathsf{x}_{M+1} - \mathsf{x}_M)$$

with the Jacobian

$$J_{ij} = \frac{\partial R_i}{\partial x_i}$$

We want $\mathbf{R}(\mathbf{x}_{M+1}) = 0$:

$$\mathbf{x}_{M+1} pprox \mathbf{x}_M - \mathbf{J}^{-1} \cdot \mathsf{R}(\mathbf{x}_M)$$

 ${\bf J}$ is approximated by a sequence of ${\bf J}_0,\ {\bf J}_1,\ {\bf J}_2,\ \dots$



Linear mixing

$$\mathbf{x}_{M+1} pprox \mathbf{x}_M - \mathbf{J}_M^{-1} \cdot \mathbf{R}(\mathbf{x}_M)$$

with

$$\mathbf{J}_{M}^{-1} = -\alpha \mathbb{1}$$

SO

$$\mathbf{x}_{M+1} = \mathbf{x}_M + \alpha \mathbf{R}(\mathbf{x}_M) = \mathbf{x}_M + \alpha (\mathbf{F}(\mathbf{x}_M) - \mathbf{x}_M)$$

SciPy

from scipy.optimize.nonlin import linearmixing



"exciting" mixing

Used in the FP-LAPW DFT code (http://exciting.sourceforge.net/)

$$\mathbf{x}_{M+1} pprox \mathbf{x}_M - \mathbf{J}_M^{-1} \cdot \mathbf{R}(\mathbf{x}_M)$$

with

$$\mathbf{J}_{M}^{-1} = -\mathrm{diag}(\beta_{1}, \beta_{2}, \beta_{3}, \dots)$$

start with $\beta_1 = \beta_2 = \beta_3 = \cdots = \alpha$ and at every iteration adjust the parameters β_i according to this very simple algorithm: if $R_i(\mathbf{x}_{M-1})R_i(\mathbf{x}_M) > 0$ then increase β_i by α otherwise set $\beta_i = \alpha$ (if $\beta_i > \alpha_{max}$, set $\beta_i = \alpha_{max}$).

SciPy

from scipy.optimize.nonlin import excitingmixing



Broyden update

The first Broyden method:

$$\mathbf{J}_{M+1} = \mathbf{J}_{M} - \frac{(\Delta \mathbf{R}(\mathbf{x}_{M}) + \mathbf{J}_{M} \cdot \Delta \mathbf{x}_{M}) \Delta \mathbf{x}_{M}^{T}}{|\Delta \mathbf{x}_{M}|^{2}}$$

The second Broyden method:

$$\mathbf{J}_{M+1}^{-1} = \mathbf{J}_{M}^{-1} + \frac{(\Delta \mathbf{x}_{M} - \mathbf{J}_{M}^{-1} \cdot \Delta \mathbf{R}(\mathbf{x}_{M})) \Delta \mathbf{R}(\mathbf{x}_{M})^{T}}{|\Delta \mathbf{R}(\mathbf{x}_{M})|^{2}}$$

starting with the linear mixing:

$$\mathbf{J}_0^{-1} = -\alpha \mathbb{1}$$

SciPy

from scipy.optimize import broyden1, broyden2



low memory second Broyden update

The second Broyden method

$$(\mathbf{J}_{M+1}^{-1} = \mathbf{J}_{M}^{-1} + rac{(\Delta \mathbf{x}_{M} - \mathbf{J}_{M}^{-1} \cdot \Delta \mathbf{R}(\mathbf{x}_{M})) \Delta \mathbf{R}(\mathbf{x}_{M})^{T}}{|\Delta \mathbf{R}(\mathbf{x}_{M})|^{2}})$$
 can be written as
$$\mathbf{J}_{M+1}^{-1} = \mathbf{J}_{M}^{-1} + \mathbf{u} \mathbf{v}^{T}$$

with

$$\mathbf{u} = \Delta \mathbf{x}_M - \mathbf{J}_M^{-1} \cdot \Delta \mathbf{R}(\mathbf{x}_M)$$
 $\mathbf{v} = \frac{\Delta \mathbf{R}(\mathbf{x}_M)}{|\Delta \mathbf{R}(\mathbf{x}_M)|^2}$

so the whole inverse Jacobian can be written as

$$\mathbf{J}_{M}^{-1} = -\alpha \mathbb{1} + \mathbf{u}_{1} \mathbf{v}_{1}^{T} + \mathbf{u}_{2} \mathbf{v}_{2}^{T} + \mathbf{u}_{3} \mathbf{v}_{3}^{T} + \cdots$$
$$\mathbf{J}_{M}^{-1} \cdot \mathbf{y} = -\alpha \mathbf{y} + \mathbf{u}_{1} (\mathbf{v}_{1}^{T} \mathbf{y}) + \mathbf{u}_{2} (\mathbf{v}_{2}^{T} \mathbf{y}) + \mathbf{u}_{3} (\mathbf{v}_{3}^{T} \mathbf{y}) + \cdots$$

SciPv

from scipy.optimize import broyden3



Other methods

The generalized Broyden method (modified Broyden method):

$$\sum_{p=M-k}^{M-1} (1 + \omega_0^2 \delta_{pn}) \Delta \mathbf{R}(\mathbf{x}_n)^T \Delta \mathbf{R}(\mathbf{x}_p) \gamma_p = \Delta \mathbf{R}(\mathbf{x}_n)^T \mathbf{R}(\mathbf{x}_M)$$

$$\mathbf{x}_{M+1} = \mathbf{x}_M + \beta_M \mathbf{R}(\mathbf{x}_M) - \sum_{p=M-k}^{M-1} \gamma_p (\Delta \mathbf{x}_p + \beta_M \Delta \mathbf{R}(\mathbf{x}_p))$$

other methods: Anderson, extended Anderson

SciPy

from scipy.optimize import broyden_generalized,
 anderson, anderson2



Finite element formulation

One particle Schrödinger equation:

$$\left(-\frac{\hbar^2}{2m}\nabla^2+V\right)\psi=E\psi.$$

FEM:

$$\begin{split} \left(K_{ij}+V_{ij}\right)q_j &= EM_{ij}q_j + F_i\,,\\ V_{ij} &= \int \phi_i V \phi_j \,\mathrm{d}V\,,\\ M_{ij} &= \int \phi_i \phi_j \,\mathrm{d}V\,,\\ K_{ij} &= \frac{\hbar^2}{2m} \int \nabla \phi_i \cdot \nabla \phi_j \,\mathrm{d}V\,,\\ F_i &= \frac{\hbar^2}{2m} \oint \frac{\mathrm{d}\psi}{\mathrm{d}n} \phi_i \,\mathrm{d}S\,. \end{split}$$

Usually we set $F_i = 0$.



SfePy

- SfePy = general finite element analysis software
- BSD open-source license
- available at
 - http://sfepy.org (developers)
 - mailing lists, issue (bug) tracking
 - we encourage and support everyone who joins!
 - http://sfepy.kme.zcu.cz (project information)
- selected applications:
 - homogenization of porous media (parallel flows in a deformable porous medium)
 - acoustic band gaps (homogenization of a strongly heterogenous elastic structure: phononic materials)
 - shape optimization in incompressible flow problems



Particle in the box

$$V(x) = egin{cases} 0, & ext{inside the box} & a imes a imes a \\ \infty, & ext{outside} \end{cases}$$

Analytic solution:

$$E_{n_1 n_2 n_3} = \frac{\pi^2}{2a^2} \left(n_1^2 + n_2^2 + n_3^2 \right)$$

where $n_i = 1, 2, 3, \ldots$ are independent quantum numbers. We chose a = 1, i.e.: $E_{111} = 14.804$, $E_{211} = E_{121} = E_{112} = 29.608$, $E_{122} = E_{212} = E_{221} = 44.413$, $E_{311} = E_{131} = E_{113} = 54.282$ $E_{222} = 59.217$, $E_{123} = E_{perm.} = 69.087$.

Numerical solution (a = 1, 24702 nodes):

(a-1, 27702 Hodes).							
	E	1	2-4	5-7	8-10	11	12-
	theory	14.804	29.608	44.413	54.282	59.217	69.087
	FEM	14.861	29.833	44.919	55.035	60.123	70.305
			29.834	44.920	55.042		70.310
			29.836	44.925	55.047		



3D Harmonic oscillator

$$V(r) = egin{cases} rac{1}{2}\omega^2 r^2, & ext{inside the box} & a imes a imes a \\ \infty, & ext{outside} \end{cases}$$

Analytic solution in the limit $a \to \infty$:

$$E_{nl} = \left(2n + l + \frac{3}{2}\right)\omega$$

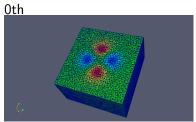
where $n, l = 0, 1, 2, \ldots$ Degeneracy is 2l + 1, so: $E_{00} = \frac{3}{2}$, triple $E_{01} = \frac{5}{2}$, $E_{10} = \frac{7}{2}$, quintuple $E_{02} = \frac{7}{2}$ triple $E_{11} = \frac{9}{2}$, quintuple $E_{12} = \frac{11}{2}$:

Numerical solution ($a=15, \omega=1, 290620 \text{ nodes}$):

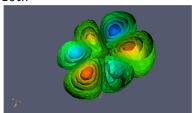
		oa. oo.ao.o (a			
E	1	2-4	5-10	11-	
theory	1.5	2.5	3.5	4.5	
FEM	1.522	2.535	3.554	4.578	
		2.536	3.555	4.579	
		2.536	3.555	4.579	
			3.555		
			3.556		
			3.556		

3D Harmonic oscillator

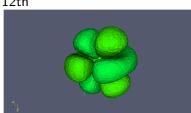
Eigenvectors:



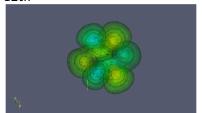
10th



12th



12th



Hydrogen atom

$$V(r) = egin{cases} -rac{1}{r}, & ext{inside the box} & a imes a imes a \\ \infty, & ext{outside} \end{cases}$$

Analytic solution in the limit $a \to \infty$:

$$E_n=-\frac{1}{2n^2}$$

where
$$n=1,2,3,\ldots$$
 Degeneracy is n^2 , so: $E_1=-\frac{1}{2}=-0.5$, $E_2=-\frac{1}{8}=-0.125$, $E_3=-\frac{1}{18}=-0.055$, $E_4=-\frac{1}{32}=-0.031$.

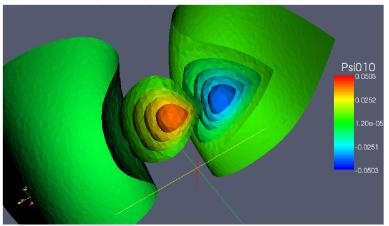
Numerical solution (a = 15, 160000 nodes):

			(-	-, -
Е	1	2-5	6-14	15-
theory	-0.5	-0.125	-0.055	-0.031
FEM	-0.481	-0.118	-0.006	



Hydrogen atom

11th eigenvalue (calculated: -0.04398532, exact: -0.056), on the mesh with 976 691 tetrahedrons and 163 666 nodes, for the hydrogen atom (V=-1/r).



FEM+DFT

We solve the Kohn-Sham equations using FEM:

$$(-\frac{1}{2}\nabla^2 + V_H(\mathbf{r}) + V_{xc}(\mathbf{r}) + v(\mathbf{r}))\psi_i(\mathbf{r}) = \epsilon_i \psi(\mathbf{r})$$

that yield the orbitals ψ_i that reproduce the density $n(\mathbf{r})$ of the original interacting system

$$n(\mathbf{r}) = \sum_{i}^{N} |\psi_{i}(\mathbf{r})|^{2}$$

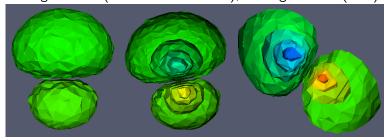
$$\nabla^{2} V_{H} = n(\mathbf{r})$$

$$v(\mathbf{r}) = \sum_{k} -\frac{Z_{k}}{|\mathbf{r} - \mathbf{R}_{k}|}$$

Current status (uniform tetrahedral mesh, 50 000 nodes):

radi	al	-6.564449519	-6.564449519	-0.3447644413	-0.3447644413	-0.1366622746
FEN	N	-3.18675417	-0.68091886	-0.65252624	-0.63762163	-0.58488204

• Bad convergence should greatly improve with a better mesh 2th eigenvector (contours and a slice), 3th eigenvector (slice)



Conclusion, Future work

- automatic mesh generation and refining
- use pseudopotentials (reduces the number of electrons to solve for)
- only depend on open source (free software) solutions

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