Some mathematical inputs in computational chemistry

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1 A priori and a posteriori Analysis for the Gross-Pitaevskii equation

- Problem presentation
- Numerical Results

2 Model Error

3 Conclusion and perspectives

Problem presentation: the Gross-Pitaevskii equation

Physical problem: Ground state of a system of bosons at very low temperature.

Two ways of seeing the problem: minimization problem - eigenvalue problem

Minimization problem: Energy functional minimization

$$I = \inf \left\{ E(v), v \in H^1_{\#}(\Omega), \int_{\Omega} v^2 = 1 \right\} \quad \text{with } \Omega = (0, 1)$$

where $E(v) = \frac{1}{2} \int_{\Omega} |\nabla v|^2 + \frac{1}{2} \int_{\Omega} V v^2 + \frac{1}{4} \int_{\Omega} v^4, \quad V \in L^p, p > 1$

Nonlinear eigenvalue problem

$$\begin{cases} \forall v \in X, (-\Delta + V + u^2)u = \lambda u \\ \int_{\Omega} u^2 = 1. \end{cases}$$

Setting: 1-Dimensional, Periodic Setting.

Remark: λ is the smallest eigenvalue and is **simple**.



Resolution method

1- Space discretization: Planewave expansion.

Expansion in Fourier series:

$$u(x) = \sum_{k} \hat{u_k} e_k(x)$$
 where $e_k(x) = e^{2\pi i k \cdot x}$

Exact space: $X = H^1_{\#}(\Omega)$. Discretized space: $X_N = \text{Span} \{e_k, |k| \leq N, k \in \mathbb{N}\}$.

$$\begin{array}{ll} \text{Discretized} & \forall v_N \in X_N, \int_{\Omega} \nabla u_N \cdot \nabla v_N + \int_{\Omega} V u_N v_N + \int_{\Omega} u_N^3 v_N - \lambda_N \int_{\Omega} u_N v_N = 0. \end{array}$$

2- Iterative resolution: Algorithm used to solve the equation on X_N :

- 1. **Initialization:** Well-chosen pair (u_N^0, λ_N^0) .
- 2. **Iterations:** Loop until convergence $(||u_N^k u_N^{k-1}||_{H^1} \text{ small})$. **Linear Problem**

$$\prod_N (-\Delta \widetilde{u_N^k} + V \widetilde{u_N^k} + (u_N^{k-1})^2 \widetilde{u_N^k}) = \lambda_N^{k-1} u_N^{k-1}.$$

Normalization $u_N^k = \widetilde{u_N^k} / \|\widetilde{u_N^k}\|_{L^2}$. Rayleigh Quotient $\lambda_N^k = \int \nabla (u_N^k)^2 + V(u_N^k)^2 + (u_N^k)^4$.

3. **Output:** Approximate eigenfunction and eigenvalue $(u_N^{k_{out}}, \lambda_N^{k_{out}})$

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Discretized
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Iterations

 (u_N^k, λ_N^k)

Error balance-Separation of error

Aim

- Analyse the error bound
- Find the origin of the error: space discretization and iterations
- Be able to refine the right parameter at each step
- Get the **best compromise** between space discretization and number of iteration that minimizes the number of computations for a given accuracy.

Two error sources:

- Size of the Fourier space 2N + 1.
- Number of iterations k.

Therefore, we decompose the main residual into two computable parts

$$R_{disc} = -\Delta u_N^k + V u_N^k + (u_N^{k-1})^2 u_N^k - \lambda_N^{k-1} u_N^{k-1}$$

$$R_{iter} = (u_N^k)^3 - (u_N^{k-1})^2 u_N^k - \lambda_N^k u_N^k + \lambda_N^{k-1} u_N^{k-1}$$

such that

$$R_N^k = R_{disc} + R_{iter}.$$

Error balance-Estimator

Writing $R_N^k = R_{disc} + R_{iter}$, we find

$$\|u - u_N^k\|_{H^1} \le \alpha(\operatorname{err}_k + \operatorname{err}_N)$$

where err_k depends on the number of iterations and

$$err_{k} = \|R_{iter}\|_{H^{-1}} + \|(V + 3(u_{N}^{k})^{2} - \lambda_{N}^{k} - 1)_{-}\|_{L^{\infty}}$$

$$\left[\frac{\|R_{iter}\|_{H^{-1}}}{\beta_{N}^{k}} + \frac{2}{\beta_{N}^{k}}|\lambda_{N}^{k} - \mu_{N}^{1}|\|u_{N}^{k} - v_{N}^{1}\|_{L^{2}}$$

$$+ \frac{3}{2}\|u_{N}^{k} - v_{N}^{1}\|_{L^{2}}^{2} \left(1 + \frac{\|2(u_{N}^{k})^{2}v_{N}^{1}\|_{H^{-1}}}{\beta_{N}^{k}}\right)\right]$$

and err_N depends on the dimension and

$$err_N = \|R_{disc}\|_{H^{-1}}$$

We note $err_{tot} \leq err_N + err_k$, with $err_{tot} = ||R_N^k||_{H^{-1}}$

These terms are computable. We use them for adaptative refinement.

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Numerical simulations: Framework

The Fourier coefficients of the potential V are given by $\hat{V}_k = -\frac{1}{\sqrt{2\pi}} \frac{1}{|k|^4 - \frac{1}{4}}$.



Exact solution

"Exact" solution

- $\bullet\,$ Calculated in a discrete space with N=500.
- Norm of the residual:

$$\|R_N^k\|_{H^{-1}} = 4.10^{-13}$$

Results no more precise than 10^{-13}



Simulations using a large dimension



Simulations with a large number of iterations



Error balance algorithm



Error balance results



Mathematical approach:



Conclusion:

- Error not guaranteed for too coarse solutions
- Error bound used for adaptive refinement
- Error guaranteed for fine solutions

Is the error guaranteed?

Practical aspects:



Conclusion:

- Error not guaranteed for too coarse solutions
- Error bound used for adaptive refinement
- Error guaranteed for fine solutions

A priori and a posteriori Analysis for the Gross-Pitaevskii equation Problem presentation

• Numerical Results



3 Conclusion and perspectives

Indeed, what is of interest for us is the solution to the full, original, Schrödinger equation. What is the link between Schrödinger and one of the feasible model.

Kohn Sham, DFT

• Hartree Fock

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post Hartree Fock methods

We are using Slatter determinants to minimize the Schrodinger energy \ldots leads to the following equation

Find $(\Phi, \lambda) \in Y \times \mathbb{R}^N$ such that, $\forall \psi \in Y$ $\frac{1}{2} \int_{\mathbb{R}^3} \nabla \varphi_i \nabla \psi_i + \int_{\mathbb{R}^3} \nabla \varphi_i \psi_i + 2 \sum_{j=1}^N \int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{|\varphi_j(y)|^2 \varphi_j(x)\psi_j(x)}{|x-y|} dx dy$ $- \sum_{j=1}^N \int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{\varphi_j(y)\varphi_j(y)\varphi_j(x)\psi_j(x)}{|x-y|} dx dy$ $= \lambda_i \int_{\mathbb{R}^3} \varphi_i \psi_i \psi_i$

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$$\begin{split} \frac{1}{2} \int_{\mathbb{R}^3} \nabla \varphi_i \nabla \psi_i + \int_{\mathbb{R}^3} V \, \varphi_i \psi_i + 2 \sum_{j=1}^N \int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{|\varphi_j(\mathbf{y})|^2 \varphi_i(\mathbf{x}) \psi_i(\mathbf{x})}{|\mathbf{x} - \mathbf{y}|} \, d\mathbf{x} \, d\mathbf{y} \\ - \sum_{j=1}^N \int_{\mathbb{R}^3} \int_{\mathbb{R}^3} \frac{\varphi_i(\mathbf{y}) \varphi_j(\mathbf{y}) \varphi_i(\mathbf{x}) \psi_i(\mathbf{x})}{|\mathbf{x} - \mathbf{y}|} \, d\mathbf{x} \, d\mathbf{y} \\ = \lambda_i \int_{\mathbb{R}^3} \varphi_i \psi_i \end{split}$$

We are using Slatter determinants...

This has led us to an eigenvalue problem ... where we have withdrawn **only** the N lowest eigenvalues : the occupied orbitals.

There are $\mathcal{N} - N$ to be used : the excited states. The basic Hartree Fock determinant is written as

$$\Psi_0(\mathbf{x}) = \frac{1}{\sqrt{N!}} \det(\varphi_i(x_j))_{i,j=1}^N$$

and we denote it as

$$\Psi_0:=\Psi[1,2,\ldots,N]$$

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The basic Hartree Fock determinant is written as

$$\Psi_0(\mathbf{x}) = \frac{1}{\sqrt{N!}} \det(\varphi_i(x_j))_{i,j=1}^N := \Psi[1, 2, \dots, N].$$

Single excited determinant can then be constructed as

$$\Psi_j^a := \Psi[1, 2, \dots, j-1, a, j+1, \dots, N]$$

where the occupied orbital *j* is replaced by the unoccupied orbital *a*. Analogously, doubly excited determinants are constructed as

$$\Psi_{j,k}^{a,b} := \Psi[1, 2, \dots, j-1, a, j+1, \dots, k-1, b, k+1, \dots, N]$$

Higher excitations involve index

$$\mu = \left(\begin{array}{ccc} \mathbf{a}_1 & \dots & \mathbf{a}_k \\ \ell_1 & \dots & \ell_k \end{array}\right)$$

where a_i designates an index of unoccupied orbital that replaces the occupied one ℓ_i associated to an excitation of order k. Such an excited determinant is denoted as $\Psi_{\mu} = X_{\mu}\Psi_0$ where X_{μ} is a k-order excitation operator.

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The Projected Coupled Cluster Method consists in the ansatz

$$\mathcal{T}=\mathcal{T}(\mathbf{t})=\sum_k T_k=\sum_\mu t_\mu X_\mu$$

e.g. the CCSD method is given by $T = T_1 + T_2 = T(\mathbf{t})$ where the unknowns are the cluster amplitudes t_{μ} that are determined by the following (nonlinear) equation

$$\forall \mu \in \mathcal{J}_1 \cup \mathcal{J}_2, \quad \mathbf{0} = < \Psi_{\mu}, e^{-T} H e^T \Psi_{\mathbf{0}} >$$

We define the residual $f_{\mu}(\mathbf{t})$ by

$$f_{\mu}(\mathbf{t}):=<\Psi_{\mu}, e^{- op}He^{ op}\Psi_{0}>$$

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$$f_{\mu}(\mathbf{t}) := <\Psi_{\mu}, e^{-T}He^{T}\Psi_{0} >$$

The natural question is then to understand the link between the norm of \mathbf{t} and the energy norm of the wave function. Following Reinhold Schneider this is provided by the quantity $\|\mathbf{t}\|_V$ defined by

$$\|\mathbf{t}\|_V^2 = \sum_{\mu \in \mathcal{J}} \varepsilon_\mu |t_\mu|^2$$

where $\varepsilon_{\mu} = \sum_{i=1}^{k} \lambda_{a_i} - \lambda_{\ell_i}$, and the λ 's are Hartree Fock eigenvalues in increasing order. This norm is equivalent to the H^1 norm of Ψ . The correct evaluation of the norm of the residual $\mathbf{f}(\mathbf{\tilde{t}}) := (f_{\mu}(\mathbf{\tilde{t}}))_{\mu}$ is thus

$$\|\mathbf{f}\|_{V'}^2 = \sum_{\mu \in \mathcal{J}} \varepsilon_{\mu}^{-1} |f_{\mu}|^2$$

Starting from an initial index set \mathcal{J}_0 composed say of single excitations. The procedure — as usual in the adaptive process — follows the rule

$\mathsf{ESTIMATE} \longrightarrow \mathsf{MARK} \longrightarrow \mathsf{REFINE}:$

at step *i*, in order to define \mathcal{J}_{i+1} we estimate those $f_{\mu}(\tilde{\mathbf{t}}_i)$ that may be added in order to improve the accuracy of the computation.

 $\forall \mu \in \mathcal{J}, \quad f_{\mu}(\mathbf{t}_i) := \langle \Psi_{\mu} | e^{-T_i} H e^{T_i} | \Psi_0 \rangle = 0$

By marking those indices associated with the those that have the largest contribution in the above V' (dual)-norm (i.e. with the relative weight ε_{μ}^{-1}) we add them in the set \mathcal{J}_i to get a \mathcal{J}_{i+1} (finer) adapted set and we continue recursively by enriching up to a level where the error estimator is small at the required accuracy.

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An adaptive strategy for Coupled Cluster Approximations



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A priori and a posteriori Analysis for the Gross-Pitaevskii equation Problem presentation

• Numerical Results

2 Model Error



Conclusion:

- A posteriori Analysis for a nonlinear eigenvalue problem
- A posteriori analysis for the error in the model

Perspectives:

- A posteriori Analysis on the Kohn-Sham model.
- Other discretizations like wavelets, finite elements.

A priori results:

 Cancès, E., Chakir, R., & Maday, Y. (2012). Numerical analysis of the planewave discretization of some orbital-free and Kohn-Sham models. ESAIM: Mathematical Modelling and Numerical Analysis, 46(02), 341-388.

A posteriori Analysis:

• Dusson, G., & Maday, Y. (2013). A Posteriori Analysis of a Non-Linear Gross-Pitaevskii type Eigenvalue Problem.

Model Error:

• in collaboration with Marco Caricato : Gaussian and Filippo Lipparini : Group of Jürgen Gauss at Mainz