FYS4480 - Quantum Mechanics for Many-Particle Systems

Project 1

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• Github repository containing programs and results:

https://github.com/evenmn/FYS4480

Abstract

The aim of this project is to study the performance of linear regression in order to fit a two dimensional polynomial to terrain data. Both Ordinary Least Square (OLS), Ridge and Lasso regression methods were implemented, and for minimizing Lasso's cost function Gradient Descent (GD) was used. A fourth method was to minimize the cost function of Ridge using GD. The fitted polynomial was visualized and compared with the data, the Mean Square Error (MSE) and R²-score were analyzed, and finally the polynomial coefficients were studied applying visualization tools and Confidence Intervals (CI). To benchmark the results, we used Scikit Learn.

We found the self-implemented OLS and Ridge regression functions to reproduce the benchmarks, and Lasso was close to reproducing the benchmark as well. However, the difference between results produced by standard Ridge regression and when minimizing its cost function is large. The OLS regression method is considered as the most successful due to its small MSE and high R²-score.

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1 Introduction

In the childhood of quantum mechanics, immense efforts were placed on Helium computations, because leading physicists found it crucial to provide calculations in agreement with experiments in order to prove the theory. The Norwegian physicist Egil Hylleraas calculated the ground state energy of the Helium atom with impressive accuracy already in 1929, which in some ways proved the quantum theory to be correct. After that, physicists and chemists have managed to study systems of ever higher complexities, thanks to better methods and stronger computers.

In this project we estimate the energy levels of the Helium and Beryllium atom, with focus on the ground state energy. For doing that, we first apply the Configuration Interaction Singles (CIS), and thereafter turn to Hartree-Fock.

2 Theory

To calculate the energies, we need to solve the Schrödinger equation

$$\hat{H} |\Psi_p\rangle = \varepsilon_p |\Psi_p\rangle \tag{1}$$

where we expect to find the exact ground state energy, ε_0 , when the correct ground state wave function (GSWF) has been used. In this project we will stick to the Born-Oppenheimer approximation, which gives the Hamiltonian when the nucleus is stationary (is not affected by the electrons),

$$\hat{H} = \sum_{i=1}^{N} t(x_i) - \sum_{i=1}^{N} k \frac{Ze^2}{r_i} + \sum_{i \le j}^{N} \frac{ke^2}{r_{ij}}.$$
 (2)

The first term gives the kinetic energy, the second gives the energy from the external potential (nucleus) and the last term gives the interaction energy. Z is the atomic number, defining the number of protons in the nucleus, k is the constant from Coulomb's law and e is the elementary change.

We now introduce the atomic units, setting $\hbar = c = e = m_e = k = 1$. The energies can be converted between atomic units and electron volts using the relation

1 a.u. =
$$2 \cdot 13.6 \text{ eV}$$
. (3)

The Hamiltonian now reads

$$\hat{H} = \hat{H}_0 + \hat{H}_I = \sum_{i=1}^N \hat{h}_0(x_i) + \sum_{i< j}^N \frac{1}{r_{ij}}$$
(4)

with $r_{ij} \equiv \frac{1}{|\mathbf{r}_i - \mathbf{r}_j|}$ and \hat{h}_0 as the one-body operator for each electron. The single particle functions (SPF) are assumed to be hydrogen-like, where the one-body energies are known from the Bohr model, stating

$$E_n = -\frac{Z^2}{2n^2} \tag{5}$$

where n is the number of nodes in the wave function. In order to calculate the two-body energies, we need to solve the integrals

$$\int r_1^2 dr_1 \int r_2^2 dr_2 \Phi_{\alpha}^*(r_1) \Phi_{\beta}^*(r_2) \frac{1}{|\boldsymbol{r}_1 - \boldsymbol{r}_2|} \Phi_{\gamma}(r_1) \Phi_{\delta}(r_2) \equiv \langle \alpha \beta | \hat{v} | \gamma \delta \rangle$$
 (6)

where the Dirac notation is used for shorthand notation. Note carefully that the Φ 's are the total wave functions, where both the radial and spin parts are included. The spin part is known to not affect the energies, and it can therefore be factorized out,

$$\langle \alpha \beta | \hat{v} | \gamma \delta \rangle = (\alpha \beta | \hat{v} | \gamma \delta) \langle \chi_{\alpha} \chi_{\beta} | \chi_{\gamma} \chi_{\delta} \rangle$$

$$= (\alpha \beta | \hat{v} | \gamma \delta) \langle \chi_{\alpha} | \chi_{\gamma} \rangle \langle \chi_{\beta} | \chi_{\delta} \rangle$$

$$= (\alpha \beta | \hat{v} | \gamma \delta) \delta_{\sigma_{\alpha} \sigma_{\gamma}} \delta_{\sigma_{\beta} \sigma_{\delta}}.$$
(7)

We observe that the integral is non-zero if and only if α and γ got the same spin, and β and δ got the same spin.

2.1 Second Quantization and particle-hole-formalism

First write some general second quantization statements, when define the Hamiltonian in second quantization

2.1.1 Wick's theorem

State Wick's theorem

2.1.2 Energy formulas

For calculating the ground state energy of an atom with atomic number Z, we need to calculate

$$\varepsilon_0 = \langle \Phi_0 | \hat{H} | \Phi_0 \rangle = \sum_i \langle i | \hat{h}_0 | i \rangle + \frac{1}{2} \sum_{ij} \left[\langle ij | \hat{v} | ij \rangle - \langle ij | \hat{v} | ji \rangle \right]$$
 (8)

$$=E_0(Z)+\dots \tag{9}$$

See appendix A. Should also add cHia and iaHja, and mention that they are gonna be used in CIS.

2.2 The Helium atom

A neutral Helium atom consists of a nucleus of two protons with two electrons orbiting, and is one of the most simple many-body systems one can study. The difficulty of dealing with many-body systems lies in the interaction, where the elements $\langle \alpha \beta | \hat{v} | \gamma \delta \rangle$ can be really tricky to handle. Hard to find exact wave function.

There exist various methods for solving this problem, where one of the most successful is to define a wave function which is varied such that the energy is minimized. This method was used by E. Hylleraas already in 1929, when he minimized the energy with a wave function of 10 variational parameters, using a mechanical desk calculator. [https://www.encyclopedia.com/science/dictionaries-thesauruses-pictures-and-press-releases/hylleraas-egil-andersen] He found the energy to be -2.90363 eV, which is close to recent experimental values. [http://www.umich.edu/chem46]

Lide 1992 Handbook of Chem. and Phys. lide 1992

2.2.1 Ground state

Helium is in its ground state when both electrons occupy the 1s orbital, thus they have opposite spins and $M_s = 0$. In this project, we will study systems of $M_s = 0$ only, such that an electron can only be excited to a state with the same spin.

Taking the Pauli principle into account, the two-body wavefunction needs to be nulled out if two particles are in the same state, and it should also be antisymmetric under exchange of two particles. We are therefore in need of a Slater Determinant (SD), given by

$$\Psi(x_1, x_2, \sigma_1, \sigma_2) = A \begin{vmatrix} \psi_{\sigma_1}(x_1) & \psi_{\sigma_1}(x_2) \\ \psi_{\sigma_2}(x_1) & \psi_{\sigma_2}(x_2) \end{vmatrix}$$
(10)

$$= A \left[\psi_{\sigma_1}(x_1) \psi_{\sigma_2}(x_2) - \psi_{\sigma_2}(x_1) \psi_{\sigma_1}(x_2) \right]$$
 (11)

where ψ_{σ_1} is the SPF with spin $m_s = \sigma_1$ and A is a normalization constant. If we assume that the SPFs are normalized, our ansatz can be written as

$$|\Phi_0\rangle = \frac{1}{\sqrt{2}}(|1\rangle|\bar{1}\rangle - |\bar{1}\rangle|1\rangle)$$
 (12)

with $|\bar{1}\rangle$ describing a particle with principle quantum number 1 and spin σ_2 , which we will illustrate with spin down. In second quantization, we can construct the ground state from the true vacuum,

$$|\Phi_0\rangle = a_1^{\dagger} a_{\bar{1}}^{\dagger} |0\rangle \,, \tag{13}$$

which is understand to take the anti-symmetry property. The ground state is illustrated in figure (1).

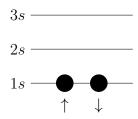


Figure 1: Ground state of Helium.

2.2.2 Excited states

When entering excited states, we can risk having angular quantum numbers unlike zero, because $l \in [0, n-1]$. This makes the computations more difficult, but in this project we will ignore that challenge setting l = 0, and thus limit us to the s-waves.

Consider now a system consisting of the orbitals 1s, 2s and 3s only. For that case, the possible energy states of the Helium atom are listed in figure (2).

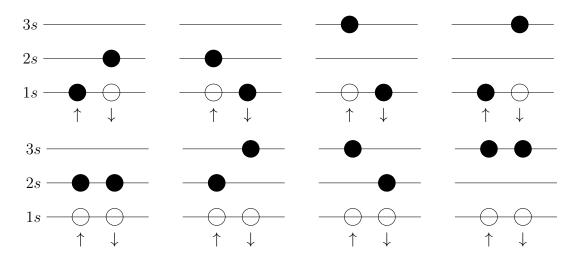


Figure 2: Possible states in the 1s, 2s and 3s orbitals of Helium.

2.3 The Beryllium atom

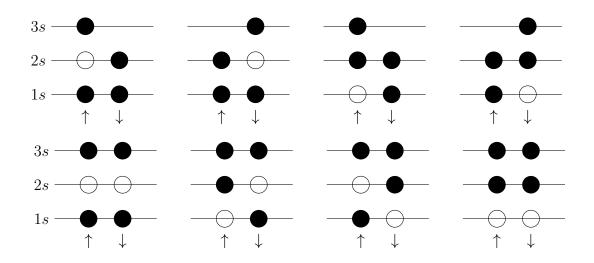


Figure 3: Possible states in the 1s, 2s and 3s orbitals of Beryllium.

3 Methods

3.1 Configuration Interaction (CI)

Here we need to show how to go from Schrodinger to the Hamilton matrix. Show some general implementations.

3.2 Hartree-Fock

The implementation could look something like this

```
def bootstrap(data, K=1000):
    dataVec = np.zeros(K)
    for k in range(K):
        dataVec[k] = np.average(np.random.choice(data, len(data)))
    Avg = np.average(dataVec)
    Var = np.var(dataVec)
    Std = np.std(dataVec)
    return Avg, Var, Std
```

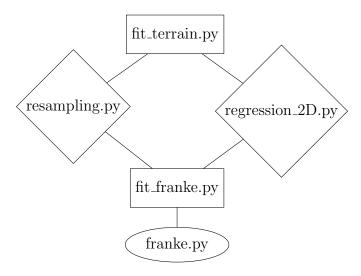


Figure 4: Code structure

4 Code

4.1 Code structure

5 Results

Table 1: Mean Square Error and R²-score presented for OLS, Ridge, Lasso and Ridge + gradient descent (RidgeGD), where noise was added to the data. The parameters used were $\lambda=1e-5$ (penalty), $\eta=1e-4$ (learning rate), niter = 1e5 (number of iterations) and $\mathcal{N}(0,\sigma^2=0.1)$ (noise). See text for more information.

| | | MSE | | | R2 | |
|---------|----------|----------|----------|--------|--------|--------|
| | Self | K-fold | Scikit | Self | K-fold | Scikit |
| | | | | | | |
| OLS | 0.008494 | 0.009119 | 0.008494 | 0.9048 | 0.8956 | 0.9048 |
| Ridge | 0.009128 | 0.009651 | 0.009128 | 0.8977 | 0.8895 | 0.8977 |
| Lasso | 0.01439 | 0.01489 | 0.01555 | 0.8387 | 0.8296 | 0.8257 |
| RidgeGD | 0.01451 | 0.01504 | 0.009128 | 0.8373 | 0.8280 | 0.8977 |

- 6 Discussion
- 7 Conclusion
- A Appendix A

Expression for energy of ground state.

B References

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