

# Exercise Ex4

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**Please use this template to submit your answers.**

If you had to modify code from the notebook, please include the modified code in your submission either as screenshot or in a

```
\begin{lstlisting}[language=Python]
\end{lstlisting}
```

environment.

You only need to include the code cells that you modified.

Note, that references to other parts of the documents aren't resolved in this template and will show as ???. Check the text of the exercises on website for the reference

## Exercise 1

Define  $S_{ij}$  using Dirac notation.

$$S_{ij} = \langle i | j \rangle = \langle \phi_i | \phi_j \rangle$$

## Exercise 2

For an orthonormal basis, what does the overlap integral array,  $\mathbf{S}$ , look like?

For an orthonormal basis,  $S_{ij} = \delta_{ij}$ .  $\mathbf{S}$  is the identity matrix. It has the dimension equals to the size of our basis set.

## Exercise 3

Use  $\mathbf{B}$  and  $\mathbf{B}_{\text{dagger}}$  and the matrix rules above to calculate the matrix  $\mathbf{S}$ .

```
1 S = B_dagger.dot(B)
```

#### Exercise 4

Describe how the notation of the `np.einsum` command correlates to the implicit summation formula written above.

```
1 S=npumpy.einsum('ip,iq->pq',B.conj(),B)
```

It means that we indicate to numpy a repeated index  $i$  on which we will make the implicit summation. Indeed we will sum on  $i$  tensor  $B_{ip}^*$  with tensor  $B_{iq}$  and reorganize the result as a new tensor  $S_{pq}$ . This corresponds to  $\sum_i B_{ip}^* B_{iq} = S_{pq}$  which is exactly the overlap matrix.

#### Exercise 5

Use the function `np.einsum()` to calculate the matrix `S`, and confirm that your answer is the same as above.

```
1 S = np.einsum('ip,iq->pq',B.conj(),B) #<your formula here>
2 print(F'S from Einstein notation:\n{S}')
```

S from Einstein notation:  $\begin{bmatrix} 1. & 0. \\ 0. & 1. \end{bmatrix}$

#### Exercise 6

Propose a different orthonormal basis, modify `phi1` and `phi2`, and verify that `S` still has the same form. There are infinitely many choices. It isn't complex... or is it?!

```
1 # two orthonormal vectors
2 phi1 = np.array([np.cos(.5), np.sin(.5)]) #insert here
3 phi2 = np.array([np.sin(.5), -np.cos(.5)]) #insert here
4
5 # construct the overlap matrix from matrix of basis vectors
6 vector_length = phi1.size #length of the vector space
7 phi1_column = phi1.reshape(vector_length,1) #this makes phi a
   column vector
8 phi2_column = phi2.reshape(vector_length,1)
9
10 # put together (concatenate) the vectors into the matrix B
11 B = np.concatenate((phi1_column,phi2_column),axis=1)
12 print(F'The matrix B:\n{B}')
```

```
13
14 B_dagger = B.conj().T
15 print(F'The matrix B^\dagger:\n{B_dagger}')
```

The matrix B:  $\begin{bmatrix} 0.87758256 & 0.47942554 \\ 0.47942554 & -0.87758256 \end{bmatrix}$   
The matrix  $B^\dagger$ :  $\begin{bmatrix} 0.87758256 & 0.47942554 \\ 0.47942554 & -0.87758256 \end{bmatrix}$   
S from Einstein notation:  $\begin{bmatrix} 1. & 0. \\ 0. & 1. \end{bmatrix}$

**Exercise 7**

How many electrons are there in total in  $H_2O$ ? How many occupied molecular orbitals would you expect?

There are 10 electrons in  $H_2O$  molecule. They are occupied orbitals 5 orbitals. We have one first orbital for the core electrons, which are paired, and are from  $O$ . Then  $O$  has two unpaired electrons which will form each a covalent bond with each electron from  $H$  atoms, leading to 1 orbital each. Finally,  $O$  has two lone pairs for which we allocate to each an orbital.

**Exercise 8**

Explain the shape (number of rows and columns) of  $S$  in terms of the AO basis set we chose.

```

1 # Construct a molecular integrals object
2 mints = psi4.core.MintsHelper(wfn.basisset())
3
4 # Overlap matrix as a psi4 Matrix object
5 S_matrix = mints.ao_overlap()
6
7 # Overlap matrix converted into an ndarray
8 S = np.asarray(S_matrix)
9
10 print(f'Shape of S is {S.shape}')
```

Shape of  $S$  is (7, 7)

Since we have 7 basis functions, we need to calculate the overlap  $S_{ij}$  for all pairs  $\{i, j\} \in \{1, \dots, 7\}^2$ . Leading to a matrix of dimension  $7 \times 7$ . STO-3G gives 5 basis functions for O ( $1s^2 2s^2 2p^4$ , one basis functions for each orbital) and 1 per H atoms ( $1s^2$ ). Therefore :

$$S = \begin{pmatrix} \langle \phi_1 | \phi_1 \rangle & \langle \phi_1 | \phi_2 \rangle & \langle \phi_1 | \phi_3 \rangle & \cdots \\ \langle \phi_2 | \phi_1 \rangle & \langle \phi_2 | \phi_2 \rangle & \langle \phi_2 | \phi_3 \rangle & \cdots \\ \langle \phi_3 | \phi_1 \rangle & \langle \phi_3 | \phi_2 \rangle & \langle \phi_3 | \phi_3 \rangle & \cdots \\ \vdots & \vdots & \vdots & \ddots \\ & & & \langle \phi_7 | \phi_7 \rangle \end{pmatrix}$$

**Exercise 9**

Based on your observations of  $S$  in the AO basis, answer the following questions

1. What do the diagonal elements of  $S$  indicate?
2. What do the off-diagonal elements of  $S$  indicate?

3. Does the Gaussian atomic orbital basis set form an orthonormal basis?

1. The diagonal terms are equal to 1, the basis sets is normalized.
2. Most of the terms are 0 but some terms are of order  $10^{-1}$  and  $10^{-2}$  indicating that the overlap between two different atomic orbitals is not always 0 and that not all states are orthogonal.
3. The Gaussian atomic orbital basis set does not form a complete orthonormal basis set due to significant overlap between some orbitals.  $S$  will have to be considered for further calculations.

#### Exercise 10

Does the result of your extra evaluation agree with what you determined previously?

```
1 isBasisOrthonormal(S)
```

Q:(T/F) The AO basis is orthonormal? A: False

False

This confirms the fact that the gaussian atomic orbital basis set does not form a complete ortho-normal basis set.

#### Exercise 11

Use the function `np.linalg.inv()` to calculate the inverse of  $S$ , and the function `splinalg.sqrtm()` to take its (matrix) square root. Execute the code below and examine the matrix  $A$ .

```
1 S_inverse = np.linalg.inv(S)
2
3 A = splinalg.sqrtm(S_inverse)
4
5 peak(A)
```

Here is a peak at the first 4 x 4 elements of the matrix:

```
[[ 1.02363458 -0.13685468 -0.00748725 0. ]
 [-0.13685468 1.15786316 0.07216005 0. ]
 [-0.00748725 0.07216005 1.03830504 0. ]
 [ 0. 0. 0. 1. ]]
```

The matrix  $A$  shows that the gaussian atomic orbital basis set does not form a complete ortho-normal basis set since it would have been the identity if this was the case. Indeed if  $S = \mathbb{I}$  then  $A = \mathbb{I}$

#### Exercise 12

What do you observe about the elements of  $A$ ? Is the matrix real or complex? Is the matrix symmetric or not?

Elements of  $A$  almost all non-zero and are positive and negative-valued.  $A$  inherits its properties from the ones of The overlap matrix  $S$  which is hermitian ( $S = S^\dagger$ ). Now  $A = S^{-1/2} = \sum_{i=1}^7 \lambda_i^{-1/2} |\lambda_i\rangle\langle\lambda_i|$ , where  $\{|\lambda_i\rangle\}$  are the eigenvalues of  $S$ , so we know that  $A$  is hermitian and therefore symmetric as shown in the figure. In our case,  $A$  is symmetric and real.

```
array([[ 1.02363458e+00, -1.36854677e-01, -7.48725467e-03,
         0.00000000e+00, -1.08457284e-18,  1.90278847e-02,
         1.90278847e-02],
       [-1.36854677e-01,  1.15786316e+00,  7.21600498e-02,
         0.00000000e+00,  3.64651649e-17, -2.2232521e-01,
        -2.2232521e-01],
       [-7.48725467e-03,  7.21600498e-02,  1.03830504e+00,
         0.00000000e+00, -3.75981753e-17, -1.18462545e-01,
        -1.18462545e-01],
       [ 0.00000000e+00,  0.00000000e+00,  0.00000000e+00,
         1.00000000e+00,  0.00000000e+00,  0.00000000e+00,
         0.00000000e+00],
       [-3.45874404e-19,  5.02534544e-17, -4.01081341e-17,
         0.00000000e+00,  1.07331473e+00,  1.75758279e-01,
        -1.75758279e-01],
       [ 1.90278847e-02, -2.2232521e-01, -1.18462545e-01,
         0.00000000e+00,  1.75758279e-01,  1.12972338e+00,
        -6.25974978e-02],
       [ 1.90278847e-02, -2.2232521e-01, -1.18462545e-01,
         0.00000000e+00, -1.75758279e-01, -6.25974978e-02,
         1.12972338e+00]])
```

Figure 1: Matrix  $A$

### Exercise 13

Use the orthogonalization matrix  $A$  to transform the overlap matrix,  $S$ . Check the transformed overlap matrix,  $S_p$ , to make sure it represents an orthonormal basis.

```
1 # Transform S with A and assign the result to S_p
2
3 S_p = A.dot(S.dot(A)) #<your equation here>
4
5 isBasisOrthonormal(S_p)
```

Q:(T/F) The AO basis is orthonormal? A: True  
True  $S'$  now represents an orthonormal basis.

### Exercise 14

The product  $A S A$  does not take the complex conjugate transpose of  $A$ . What conditions (properties of  $A$ ) make that ok?

$A$  is hermitian and real, making it a symmetric matrix, indeed we have :  $A^\dagger = A$ .

### Exercise 15

Based on the definition of  $C'$ , propose a definition of  $C$  in terms of  $A$  and  $C'$ . Justify your equation.

$$C' = A^{-1}C$$

$$AC' = AA^{-1}C$$

$$C = AC' \quad \text{where we used that } A \text{ is invertible such that } A^{-1}A = \mathbb{I}$$

### Exercise 16

In the cell below, use the core Hamiltonian matrix as your initial guess for the Fock matrix. Transform it with the same A matrix you used above. To calculate the eigenvalues, `vals`, and eigenvectors, `vecs`, of matrix M using `vals, vecs = np.linalg.eigh(M)`.

```
1 # Guess for the Fock matrix
2 F = H # Replace
3
4 # Transformed Fock matrix
5 F_p = A.dot(F.dot(A)) # Replace
6
7 # Diagonalize F_p for eigenvalues & eigenvectors with NumPy
8 e, C_p = np.linalg.eigh(F_p) # Replace
```

### Exercise 17

Display, i.e., `print`, the coefficient matrix and confirm it the correct size

```
1 print(C_p)
2 print("The size of the coefficient matrix is:", C_p.shape)
```

```
[[-9.93397459e-01  1.06813566e-01 -1.35307498e-16 -4.17010426e-02
  1.76149298e-15 -3.65697236e-03  1.33010607e-16]
 [-1.14180122e-01 -9.13965568e-01  1.27788258e-15  3.67704764e-01
 -1.55859911e-14  1.28152437e-01 -2.37455910e-15]
 [-7.66151195e-04 -3.68538378e-01 -2.99255600e-15 -9.28905590e-01
  4.00351377e-14  3.62392157e-02 -1.59671447e-15]
 [ 0.00000000e+00 -3.33066907e-16  3.57353036e-16 -4.30211422e-14
 -1.00000000e+00 -3.33066907e-16  5.55111512e-17]
 [ 2.31531739e-18  4.71826794e-17  9.98991226e-01 -3.20984454e-15
  4.77769914e-16 -1.24429917e-15 -4.49058022e-02]
 [-7.86768545e-03 -9.33736227e-02 -3.17531973e-02  9.71167846e-03
 -3.92055707e-16 -7.00803217e-01 -7.06393470e-01]
 [-7.86768545e-03 -9.33736227e-02  3.17531973e-02  9.71167846e-03
  2.74078108e-16 -7.00803217e-01  7.06393470e-01]]
```

The size of the coefficient matrix is: (7, 7)

### Exercise 18

Use A and the formula you proposed previously to transform the coefficient matrix back to the AO basis. Confirm that the resulting matrix appears reasonable, i.e., similar size and magnitude

```
1 # Transform the coefficient matrix back into AO basis
2 # Transform the coefficient matrix back into AO basis
3 C = A.dot(C_p)
4 print(C)
```

```

5 print("\nThe size of C is :",C.shape)
6 print("Frobenius norm of C to check magnitude : ||C||_F = Tr[C^T C]
    =", np.trace((C.T).dot(C)))

```

```

[[-1.00154358e+00  2.33624458e-01 -3.02733347e-16 -8.56842079e-02
  3.63414278e-15 -4.82226017e-02  9.45116473e-16]
 [ 7.18933084e-03 -1.05793876e+00  1.40864302e-15  3.60110465e-01
 -1.53723457e-14  4.63121174e-01 -9.15741996e-15]
 [ 2.67129883e-04 -4.27284243e-01 -3.00096682e-15 -9.39942475e-01
  4.04447865e-14  2.12940092e-01 -5.11420285e-15]
 [ 0.00000000e+00 -3.33066907e-16  3.57353036e-16 -4.30211422e-14
 -1.00000000e+00 -3.33066907e-16  5.55111512e-17]
 [-2.15774773e-17  6.88978035e-17  1.06107023e+00 -3.15567076e-15
  3.95718955e-16 -6.17464044e-15 -2.96507060e-01]
 [-1.82134850e-03  1.49253289e-01  1.37720978e-01  3.78578935e-02
 -1.61997318e-15 -7.80700283e-01 -8.50140251e-01]
 [-1.82134850e-03  1.49253289e-01 -1.37720978e-01  3.78578935e-02
 -9.93671961e-16 -7.80700283e-01  8.50140251e-01]]

```

The size of C is : (7, 7)

Frobenius norm of C to check magnitude :  $\|C\|_F = \text{Tr}[C^T C] = 2.933564513151939$

### Exercise 19

Build the density matrix, D, from the occupied orbitals, C\_occ, using the function np.einsum(). **Hint** Look at (??)

```

1 # Build density matrix from occupied orbitals
2 D = np.array(np.einsum('pi,qi->pq',C_occ.conj(),C_occ)) # replace
    this command with np.einsum() as described above
3 print(f'The shape of D is {D.shape}')

```

The shape of D is (7, 7)

### Exercise 20

Define J in terms of the density matrix, D, and the electron repulsion integral tensor, I, using np.einsum().

```

1 #Define J
2 J = np.einsum('rs,pqrs->pq',D,I) # use np.einsum here

```

### Exercise 21

Define K in terms of the density matrix, D, and the electron repulsion integral tensor, I, using einsum().

```

1 #Define K
2 K = np.einsum('rq,pqrs->ps',D,I) # use np.einsum here

```

### Exercise 22

Define  $F$  in terms of  $H$ ,  $J$ , and  $K$ . (Recall The Hartree-Fock procedure)

```
1 #Define F as a function of H, J, and K
2 F = H + 2*J - K # insert code here
```

### Exercise 23

Calculate the SCF energy based on  $H$ ,  $F$ , and  $D$  using `np.einsum()`.

```
1 E_nuc = mol.nuclear_repulsion_energy()
2 SCF_E = E_nuc + np.einsum('pq,pq->', H + F, D) # insert your code
   here, don't forget to use E_nuc
3 print(F'Energy = {SCF_E:.8f}')
```

Energy = -73.28579644

### Exercise 24

Based on the result of the calculation in ??, is this a reasonable answer?

We found previously :

The Hartree-Fock ground state energy of the water is: -74.94207989868096 Eh

The relative error is :  $e = \frac{-73.285780 + 74.94208}{74.94208} \approx 2.21\%$ , which is significant. This means that our initial guess of the Fock matrix being the hamiltonian  $H$  leads to an SCF energy that is not a reasonable answer.

### Exercise 25

Describe a procedure (i.e. identify the steps) that will update coefficients and compute a new density matrix based on the updated values of the Fock matrix.

We obtained the new Fock matrix  $F = H + 2J - K$ . We should now take  $F' = AFA$ , find its eigenvalues and eigenvectors which give the coefficient matrix  $C'$  and then  $C = AC'$  which are the coefficient that we want to optimize in order to minimize SCF Energy. From that we can reevaluate the important quantities such as the electron density matrix  $D$  which is now refined compared to before and recalculate SCF energy to check for convergence. Each iteration, the energy decreases by taking the last  $F$  since the new one incorporates better electron interactions.

### Exercise 26

Using the procedure proposed above, calculate the updated coefficients

```
1 # ==> Nuclear Repulsion Energy <==
2 E_nuc = mol.nuclear_repulsion_energy()
3
4 # ==> SCF Iterations <==
```



```

5 # Pre-iteration energy declarations
6 SCF_E = 0.0
7 E_old = 0.0
8
9 # ==> Set default program options <==
10 # We continue recalculating the energy until it converges to the
    level we specify.
11 # The variable 'E_conv' is where we set this level of convergence.
    We also set a
12 # maximum number of iterations so that if our calculation does not
    converge, it
13 # eventually stops and lets us know that it did not converge.
14 # Maximum SCF iterations
15 MAXITER = 40
16 # Energy convergence criterion
17 E_conv = 1.0e-6
18
19 print('==> Starting SCF Iterations <==\n')
20
21 # Begin Iterations
22 for scf_iter in range(1, MAXITER + 1):
23
24     # Build Fock matrix (meaning define J, K with density matrix
        and then define F with J and k)
25
26     # <your code here>
27     J = np.einsum('rs,pqrs->pq',D.conj(),I)
28     K = np.einsum('rq,pqrs->ps',D.conj(),I)
29     F = H + 2*J - K
30
31
32     # Compute SCF energy (don't forget E_nuc!)
33
34     SCF_E = E_nuc + np.einsum('pq,pq->',H + F,D) #<your formula
        here>
35
36     print(F'SCF Iteration {scf_iter}: Energy = {SCF_E:.8f} dE = {
        SCF_E - E_old:.8f}')
37
38     # Check to see if the energy is converged. If it is break out
        of the loop.
39     # If it is not, set the current energy E_old
40
41     if (abs(SCF_E - E_old) < E_conv):
42         break
43     E_old = SCF_E
44
45     # Compute new coefficient & density matrices (Exercise 26)
46
47     # <your code here>
48     F_p = A.dot(F.dot(A))
49
50     # find e and C_p
51     e, C_p = np.linalg.eigh(F_p)
52
53     # find C
54     C = A.dot(C_p)

```

```

55
56     # Select the occupied MOs
57     C_occ = C[:, :ndocc]
58
59     # Update the density matrix
60     D = np.einsum('pi,qi->pq', C_occ.conj(), C_occ)
61
62     # MAXITER exceeded?
63     if (scf_iter == MAXITER):
64         psi4.core.clean()
65         raise Exception("Maximum number of SCF iterations exceeded.
66     ")
67
68 # Post iterations
69 print('\nSCF converged.')
70 print(F'Final RHF Energy: {SCF_E:.6f} [Eh]')

```

---

==> Starting SCF Iterations <==

```

SCF Iteration 1: Energy = -74.82812526 dE = -74.82812526
SCF Iteration 2: Energy = -74.93548796 dE = -0.10736270
SCF Iteration 3: Energy = -74.94147771 dE = -0.00598975
SCF Iteration 4: Energy = -74.94197197 dE = -0.00049425
SCF Iteration 5: Energy = -74.94205603 dE = -0.00008407
SCF Iteration 6: Energy = -74.94207439 dE = -0.00001836
SCF Iteration 7: Energy = -74.94207862 dE = -0.00000423
SCF Iteration 8: Energy = -74.94207960 dE = -0.00000098

```

SCF converged.

Final RHF Energy: -74.942080 [Eh]

```

1 # ==> Compare our SCF to Psi4 <==
2 # Call psi4.energy() to compute the SCF energy
3 SCF_E_psi = psi4.energy('SCF')
4 psi4.core.clean()
5
6 print(f"PSI4 {SCF_E_psi:.6f} Ours {SCF_E:.6f} Absolute
7     difference {np.abs(SCF_E-SCF_E_psi)}")

```

PSI4 -74.942080 Ours -74.942080 Absolute difference 2.9805295298501733e-07

### Bonus Exercise 27

Modify the value of E\_conv and describe its effect the number of iterations. Provide your code, the output of the SCF calculation and the number of iterations.

```

1 E_conv_values = [1e-3, 5e-4, 1e-4, 5e-5, 1e-6, 5e-7, 1e-8, 5e-9, 1e
2     -10, 1e-12]

```

```

3 iterations_per_econv = []
4 abs_diffs_per_econv = []
5
6 for E_conv in E_conv_values:
7     print(f"\n\n==== Running SCF with E_conv = {E_conv:.1e} =====\n")
8
9     # We restart the whole process
10    # Guess for the Fock matrix
11    F = H # Replace
12
13    # Transformed Fock matrix
14    F_p = A.dot(F.dot(A)) # Replace
15
16    # Diagonalize F_p for eigenvalues & eigenvectors with NumPy
17    e, C_p = np.linalg.eigh(F_p) # Replace
18
19    # Retrieve C
20    C = A.dot(C_p)
21
22    C_occ = C[:, :ndocc]
23
24    # Build density matrix from occupied orbitals
25    D = np.array(np.einsum('pi,qi->pq',C_occ.conj(),C_occ))
26
27    # Build electron repulsion integral (ERI) Tensor
28    I = np.asarray(mints.ao_eri())
29
30    #Define J
31    J = np.einsum('rs,pqrs->pq',D.conj(),I) # use np.einsum here
32
33    #Define K
34    K = np.einsum('rq,pqrs->ps',D.conj(),I) # use np.einsum here
35
36    #Define F as a function of H, J, and K
37    F = H + 2*J - K # insert code here
38
39    E_nuc = mol.nuclear_repulsion_energy()
40    SCF_E = E_nuc + np.einsum('pq,pq->',H + F,D)
41
42    # insert remaining steps here, you should end with a new C
43    # transform F to the orthonormal basis
44    F_p = A.dot(F.dot(A))
45
46    # find e and C_p
47    e, C_p = np.linalg.eigh(F_p)
48
49    # find C
50    C = A.dot(C_p)
51
52    # Select the occupied MOs
53    C_occ = C[:, :ndocc]
54
55
56    # Update the density matrix
57    D = np.einsum('pi,qi->pq', C_occ.conj(), C_occ)
58

```

```

59 # ==> Nuclear Repulsion Energy <==
60 E_nuc = mol.nuclear_repulsion_energy()
61
62 # ==> SCF Iterations <==
63 # Pre-iteration energy declarations
64 SCF_E = 0.0
65 E_old = 0.0
66
67 # ==> Set default program options <==
68 # Maximum SCF iterations
69 MAXITER = 40
70 # Energy convergence criterion comes from the loop variable
71 E_conv
72
73 print('==> Starting SCF Iterations <==\n')
74
75 # Begin Iterations
76 for scf_iter in range(1, MAXITER + 1):
77     # Build Fock matrix
78     J = np.einsum('rs,pqrs->pq',D.conj(),I)
79     K = np.einsum('rq,pqrs->ps',D.conj(),I)
80     F = H + 2*J - K
81
82     # Compute SCF energy (don't forget E_nuc!)
83     SCF_E = E_nuc + np.einsum('pq,pq->',H + F,D)
84
85     print(F'SCF Iteration {scf_iter}: Energy = {SCF_E:.8f} dE =
86           {SCF_E - E_old:.8f}')
87
88     # Convergence check
89     if (abs(SCF_E - E_old) < E_conv):
90         iter_count = scf_iter # record how many iterations
91         this run took
92         break
93         E_old = SCF_E
94
95     # New coefficients & density
96     F_p = A.dot(F.dot(A))
97     e, C_p = np.linalg.eigh(F_p)
98     C = A.dot(C_p)
99     C_occ = C[:, :ndocc]
100     D = np.einsum('pi,qi->pq', C_occ.conj(), C_occ)
101
102     # MAXITER exceeded?
103     if (scf_iter == MAXITER):
104         psi4.core.clean()
105         raise Exception("Maximum number of SCF iterations
106 exceeded.")
107
108 # Post iterations
109 print('\nSCF converged.')
110 print(F'Final RHF Energy: {SCF_E:.6f} [Eh]')
111
112 # ==> Compare our SCF to Psi4 <==
113 SCF_E_psi = psi4.energy('SCF')
114 psi4.core.clean()

```

```

112     abs_diff = np.abs(SCF_E - SCF_E_psi)
113     print(f"PSI4 {SCF_E_psi:.6f} Ours {SCF_E:.6f} Absolute
114           difference {abs_diff}")
115
116     # Save metrics for this tolerance
117     iterations_per_econv.append(iter_count)
118     abs_diffs_per_econv.append(abs_diff)
119
120 # === Plots: side-by-side ===
121 fig, axes = plt.subplots(1, 2, figsize=(12, 4), constrained_layout=
122                        True)
123
124 # Left: iterations vs E_conv
125 axes[0].plot(E_conv_values, iterations_per_econv, marker='o', color
126             ="b")
127 axes[0].set_xscale('log')
128 axes[0].set_xlabel('E_conv')
129 axes[0].set_ylabel('Number of SCF iterations')
130 axes[0].set_title('Iterations vs. convergence tolerance')
131 axes[0].grid(True)
132
133 # Right: |SCF_E - SCF_E_psi| vs E_conv
134 axes[1].plot(E_conv_values, abs_diffs_per_econv, marker='o', color =
135             "orange")
136 axes[1].set_xscale('log')
137 axes[1].set_xlabel('E_conv')
138 axes[1].set_ylabel('|SCF_E - SCF_E_psi| [Eh]')
139 axes[1].set_title('Absolute energy difference vs. tolerance')
140 axes[1].grid(True)
141
142 plt.show()

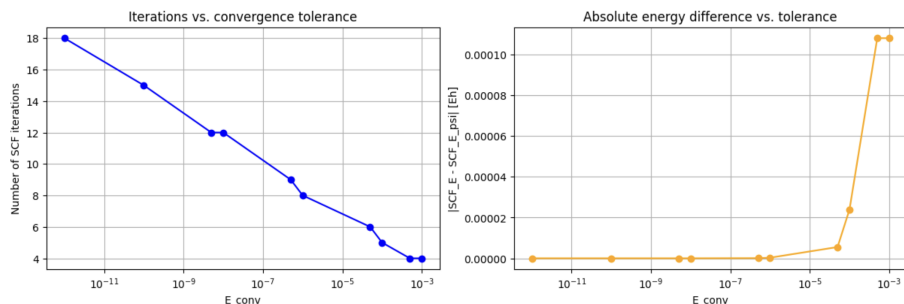
```

Here is the track of the SCF iterations for the following values :

```

1 E_conv_values = [1e-3, 5e-4, 1e-4, 5e-5, 1e-6, 5e-7, 1e-8, 5e-9, 1e
2                  -10, 1e-12]

```



We observe that decreasing the threshold  $E_{conv}$  leads to higher accuracy (min absolute error of  $3.13 \cdot 10^{-13}$ ) and longer computation time (18 iterations). Therefore  $E_{conv}$  plays a substantial role in terms of accuracy and convergence time, we have to make a trade-off between the two. For such small molecules we don't have much problems but for more complicated molecules we can reasonably extend this notion of trade-off.

Figure 2: SCF iterations for different values of  $E_{conv}$

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