

Assignment 3

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Handin: Oct. 6th. 12am

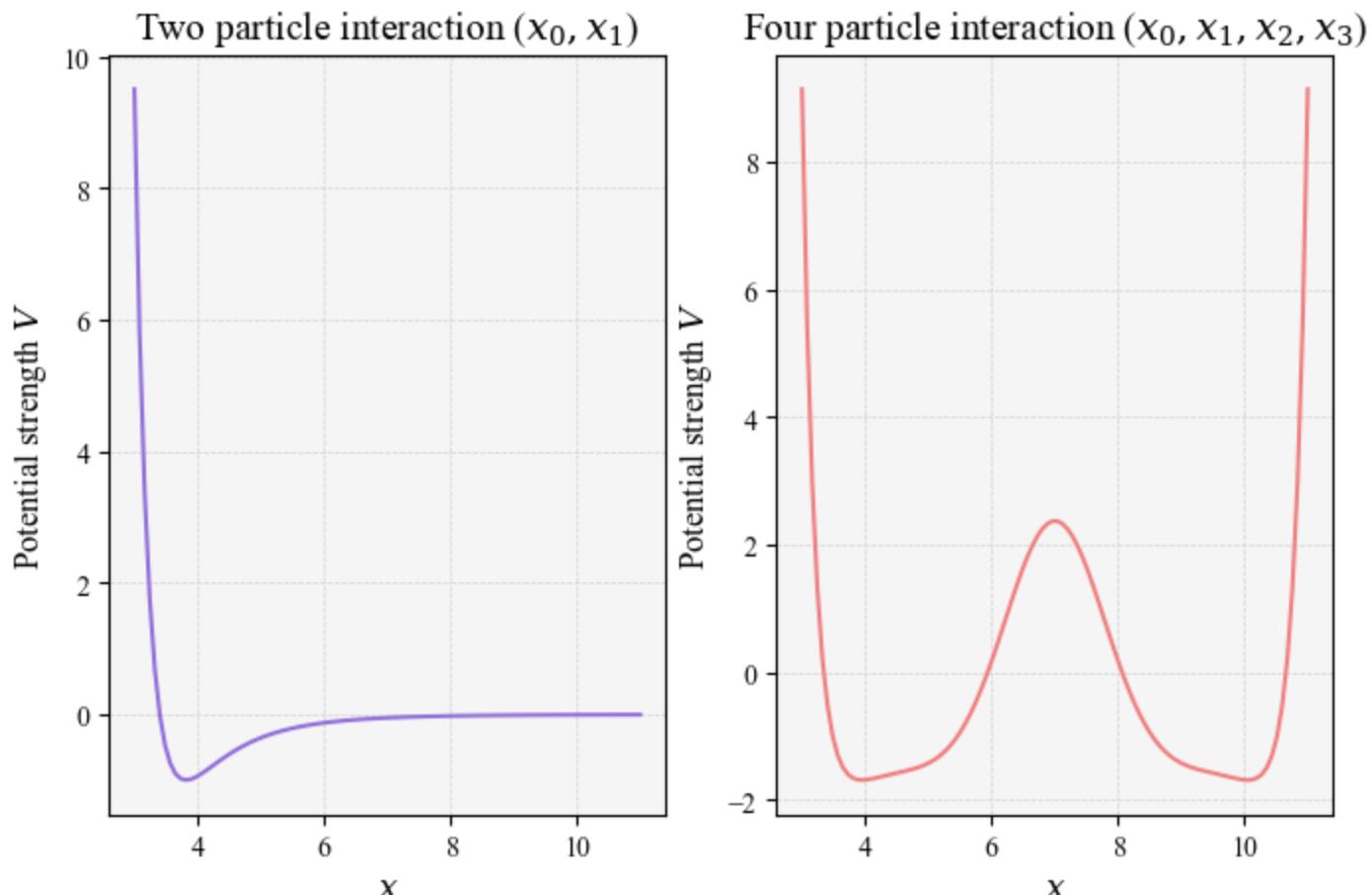
I made two PDFs for this assignment:

1. (assignment3_WMC.pdf) A PDF including all of the functions described in the exercises along with all plots, markdown and printouts.
2. (assignment3.pdf) A PDF where most of the code is hidden including the functions mentioned in the assignment text.
Printouts, markdown and figures as left as the only answer for the exercises.

Please note in the assignment comments which one is the best for you to correct.

Questions for Week 4: Solving Nonlinear equations

(A1) + (A2) Potentials for N=2 and N=4 particles



The two plots shown above are plots of the LJ-potential for a two particle system and a four particle system. The only variable in the system is the x-position of particle x_0 hence the x-axis.

The convergence test described in the assignment is implicit in all the functions, and all printouts show the root obtained from the specific algorithm along with the number of function-calls it took to converge towards that solution.

(B) Bisection root function

Root found at $x=3.401$ in the space $x = [2, 6]$ with $n=48$ function calls
Is root of x similar to σ (3.401)? True

(C) Newton Raphson solver

```
Root found at x=3.401 starting from x0=2 with n=26 function calls
Is root of x similar to σ (3.401)? True
```

(D) NR solver and bisection root function combination

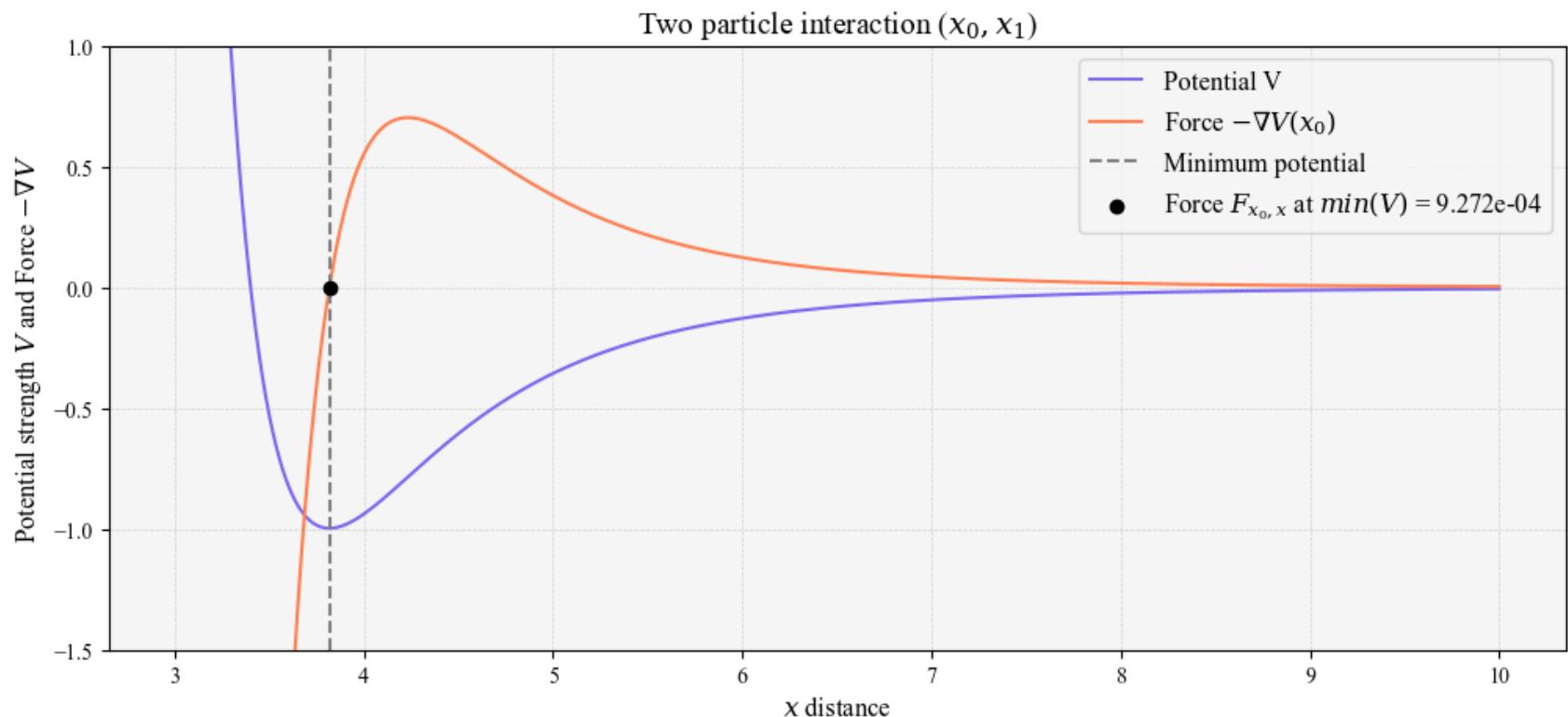
```
Root found at x=3.401 with n=40 function calls
Is root of x similar to σ (3.401)? True
```

(E)

```
Force working on particle x0 at x = 3.0: [-54.9536532  0.          0.          ]
Force working on particle x1 at x = 3.0: [54.9536532  0.          0.          ]
```

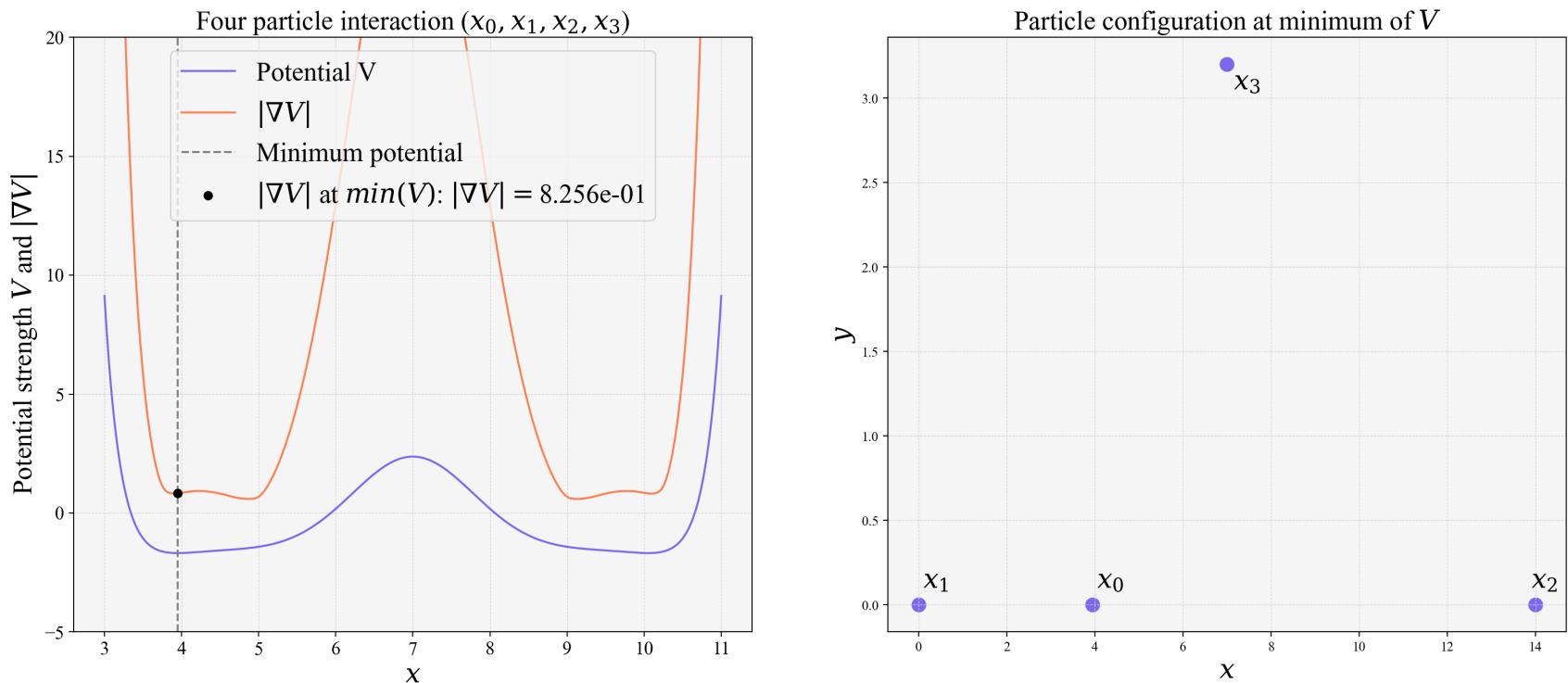
The negative gradient of the potential is the force acting on the system, and the direction of the force. So each element of the gradient tells us the force working on each particle in x , y and z . As we are working with two particles laying in the x plane the only force component between them is in the x direction. According to newtons third law, an object x_0 acting with a force on x_1 experience an equal and opposite force on itself. This is why the two quantities are equal and opposite and only have components in the x direction.

```
Out[8]: (-1.5, 1.0)
```



At the minimum of the potential, the force on the two particles is exactly zero, so the particles are in an a rest position (equilibrium).

```
Force working on particle x0 at x = 3.950: [-55.21445712 -0.20712564 0.          ]
Force working on particle x1 at x = 3.950: [ 5.49325687e+01 -9.47811027e-03 0.00000000e+00]
Force working on particle x2 at x = 3.950: [ 0.02298137 -0.00947811 0.          ]
Force working on particle x3 at x = 3.950: [0.25890705 0.22608186 0.          ]
```



In the four particle system, the magnitude of the force is never zero (yellow line in the plot). If the force is split up into x and y components (the force in z is 0), the force in one direction might be zero for a certain x , but is never zero in both directions at the same time. I believe this is due to the assymetry of the system shown in the second plot above.

(F)

Optimal step length alpha: 0.452 with n=31 function calls
 Gradient at $(X_0 + \text{alpha} * d)$ in direction d: 1.1624196055081193e-08

Questions for week 5: Nonlinear Optimization

(G) Golden section minimum

```
In [27]: a, b = 0, 1
X0 = np.array([[4, 0, 0], [0, 0, 0], [14, 0, 0], [7, 3.2, 0]])
x_opt, n_calls = golden_section_min(line_V(LJhelp.V, X0, d), a, b)
print(f"Optimal x: {x_opt:.3f} with n={n_calls} function calls")
print("Is x similar to alpha from linesearch (0.452)?", np.isclose(x_opt, alpha, atol=1e-2))
```

Optimal x: 0.452 with n=36 function calls
 Is x similar to alpha from linesearch (0.452)? True

Below the optimal distance r_0 is found for the two particle potential using the Golden-Section-minimizer

```
In [ ]: r0_res, nc0 = golden_section_min(V_two, 2, 6)
print(r"r0 for two particle system: " + f"{r0_res:.3f} with n={nc0} function calls")
```

r0 for two particle system: 3.817 with n=39 function calls

(H) BFGS minimizer

```
In [16]: X2 = ArStart['Xstart2']
X_opt, n_calls, converged = BFGS(LJhelp.flat_V, LJhelp.flat_gradV, X2, verbose=True)
X_opt = X_opt.reshape(-1, 3)
print(f"\nOptimized configuration:\n{X_opt} \nfor N={len(X_opt)} particles")
distances = LJhelp.distance(X_opt)
off_diag = [distances[i, j] for i in range(len(distances)) for j in range(len(distances)) if i != j]

print(f"\nInter-particle distances:\n{distances}\n")
print(f"The inter-particle distance is equivalent to the result from (g): {np.allclose(off_diag, r0_res)}")
```

Converged after 6 iterations and 13 function calls.

Optimized configuration:
[[3.68186699 3.93837516 1.48584684]
 [3.00028518 0.30055214 0.55046137]]
for N=2 particles

Inter-particle distances:
[[0. 3.81749343]
 [3.81749343 0.]]

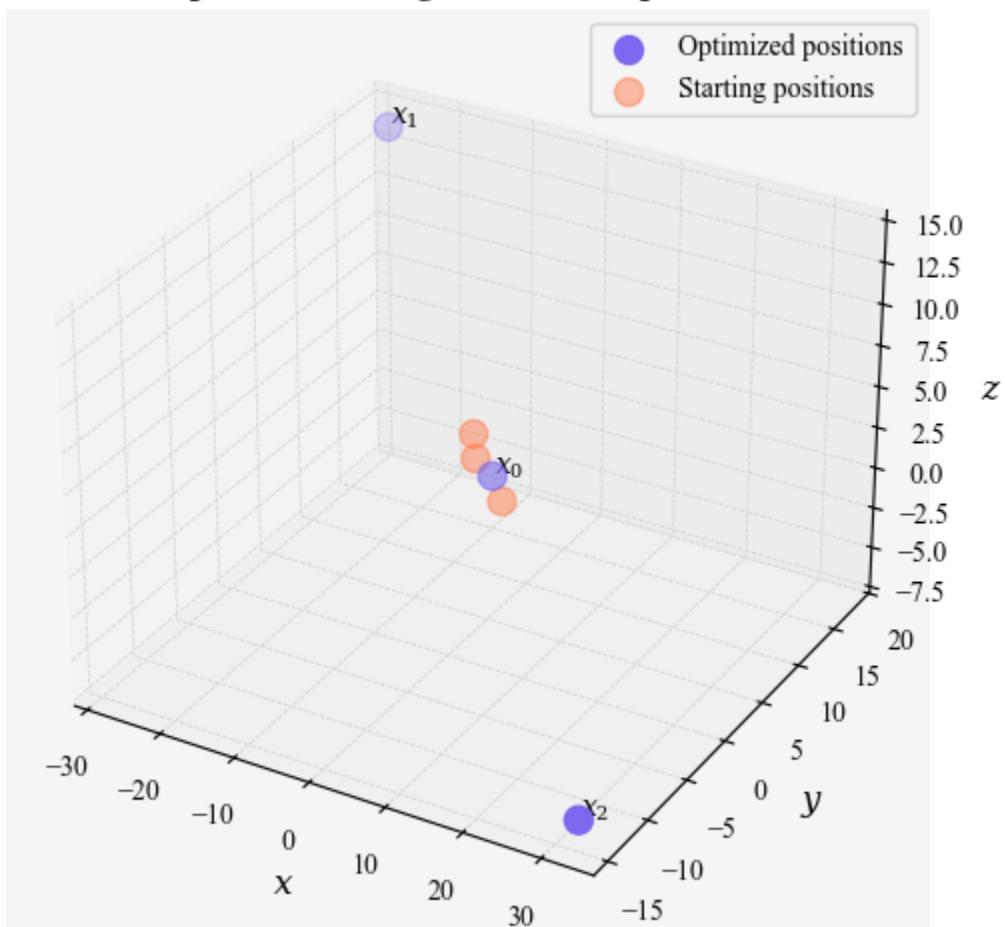
The inter-particle distance is equivalent to the result from (g): True

(I) BFGS minimizer for different N

```
In [20]: N_spec = [3, 20]
bfgs_df = bfgs_plot(N_spec=N_spec)
print("Highest N with convergence:", max_n)
print("BFGS optimization results:\n")
display(bfgs_df.style.hide(axis="index"))
```

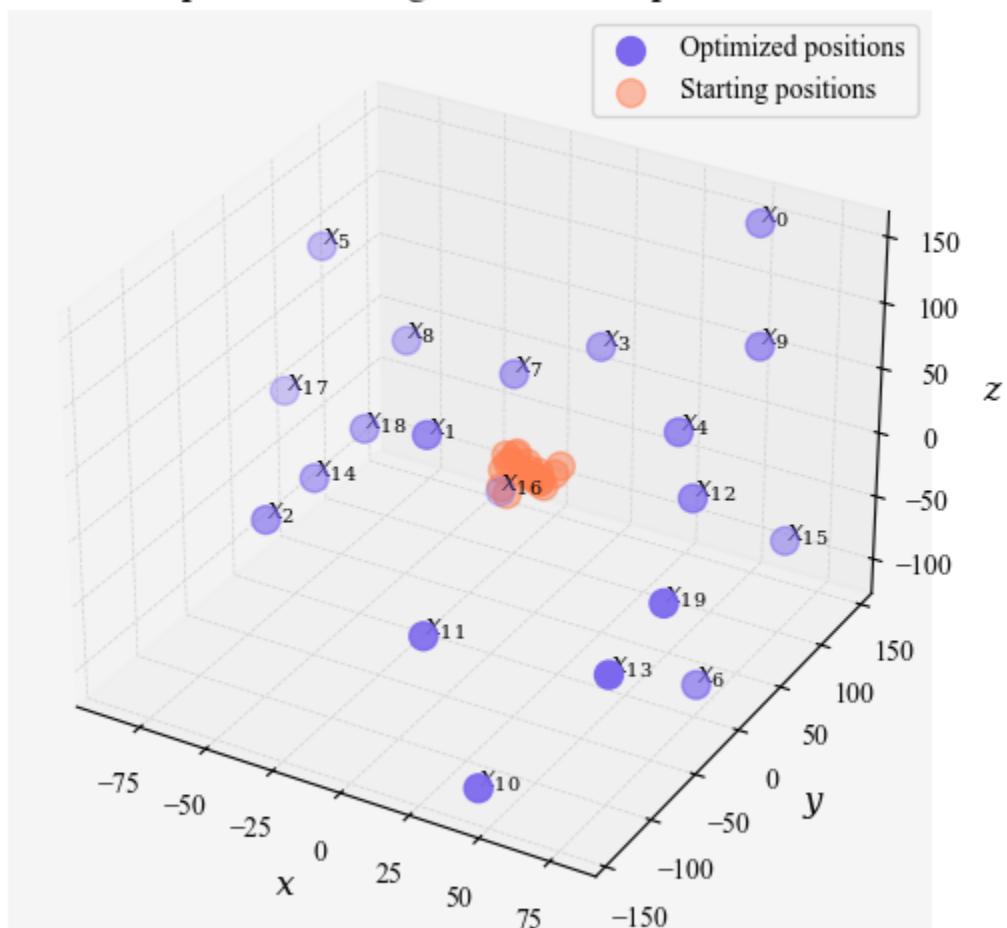
Figure

Optimized configuration for 3 particles



Figure

Optimized configuration for 20 particles



Highest N with convergence: 20
BFGS optimization results:

| N | Function calls | Total # of bonds | Bonds within 1% of r0 | Converged |
|----|----------------|------------------|-----------------------|-----------|
| 2 | 13 | 1 | 1 | True |
| 3 | 33 | 3 | 0 | True |
| 4 | 39 | 6 | 0 | True |
| 5 | 101 | 10 | 0 | True |
| 6 | 59 | 15 | 0 | True |
| 7 | 87 | 21 | 0 | True |
| 8 | 51 | 28 | 0 | True |
| 9 | 7 | 36 | 0 | True |
| 20 | 61 | 190 | 0 | True |

From the table and the plots, it is evident, that even though all configurations converged, all of them (except for $N = 2$) failed to have any Van der Waals bonds or even form the disired lattice structure.

(G) Line-search BFGS algorithm

```
In [23]: max_n = max(conv_n) if len(conv_n) > 0 else None
print("Highest N with convergence:", max_n)
print("Summary of line-search BFGS results:")
display(ls_bfgs_df.style.hide(axis="index"))
```

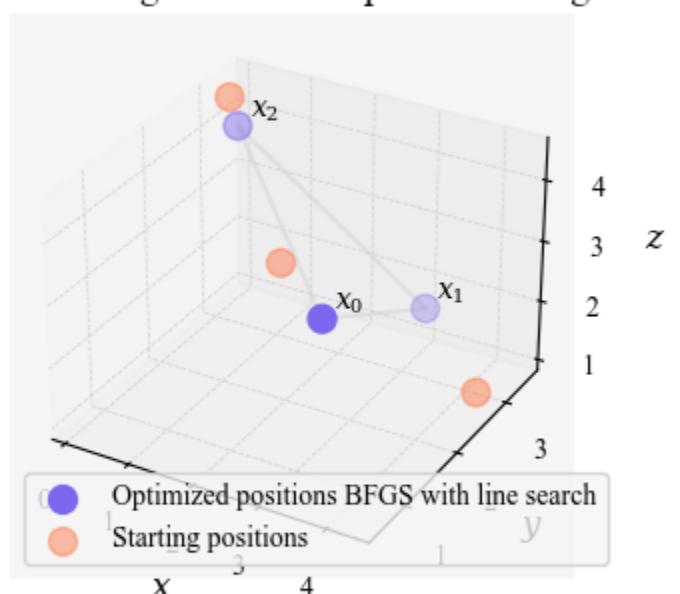
```
Highest N with convergence: 20
Summary of line-search BFGS results:
```

| N | Function calls | Total # Bonds | Bonds within 1% of r0 | converged |
|----|----------------|---------------|-----------------------|-----------|
| 2 | 39 | 1 | 1 | True |
| 3 | 647 | 3 | 3 | True |
| 4 | 761 | 6 | 6 | True |
| 5 | 24321 | 10 | 9 | True |
| 6 | 8741 | 15 | 12 | True |
| 7 | 6613 | 21 | 15 | True |
| 8 | 5169 | 28 | 18 | True |
| 9 | 2965 | 36 | 19 | True |
| 20 | 66957 | 190 | 23 | True |

As seen from the table, many more Van der Waals bonds are achieved between the particles. This is also visible in the plots below, where most configurations form nice lattice-like structures. This came at the cost of a lot more function calls, but the algorithms are generally fast to run.

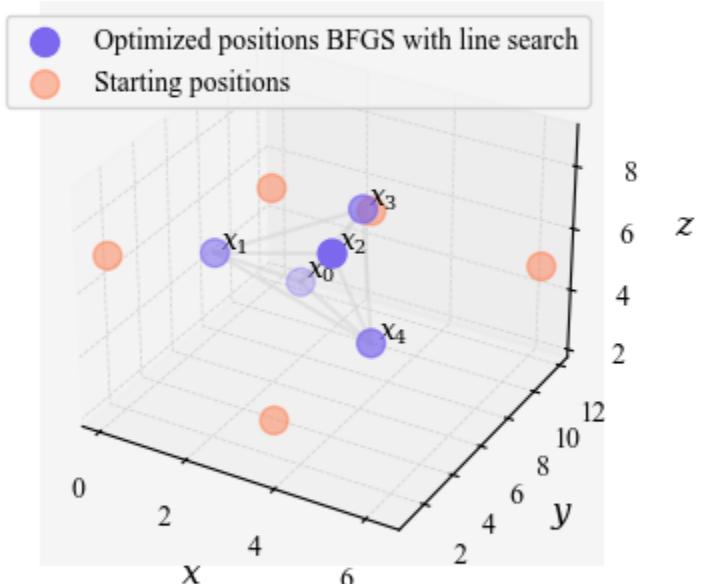
Figure

Optimized configuration for 3 particles using BFGS



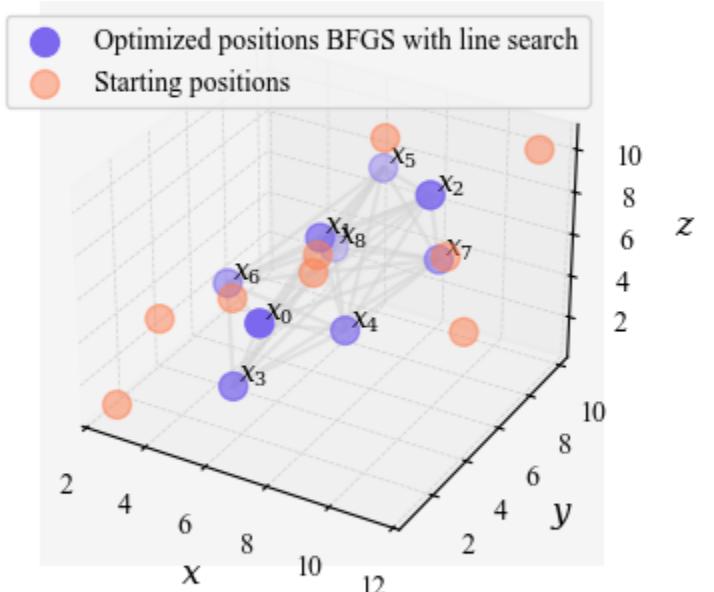
Figure

Optimized configuration for 5 particles using BFGS



Figure

Optimized configuration for 9 particles using BFGS



The plots above show the optimized particle positions from the BFGS algorithm with implemented line searching. For the three chosen $N \in [3, 5, 9]$ the particles behave as expected forming a lattice in 3D where most of the bonds are within 1% of r_0 . For the $N = 20$ configuration, all except one particle behaves well - one is placed very far from all other particles.