

Lab 5: Monte Carlo

SI1136 Simulation and Modeling

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5.1

5.2

5.3

5.4

Appendix

5.1

5.1 Project

Use the Metropolis method to calculate

$$\langle x \rangle = \frac{\int_0^\infty x e^{-x} dx}{\int_0^\infty e^{-x} dx} . \quad (1)$$

Try different, also sufficiently large values of the parameter delta. Study σ/N and compare with the actual difference to the exact answer.¹

¹Note: N_0 is not critical when $x_0 = 0$.

5.1 Background

Exact value of the integral (1).

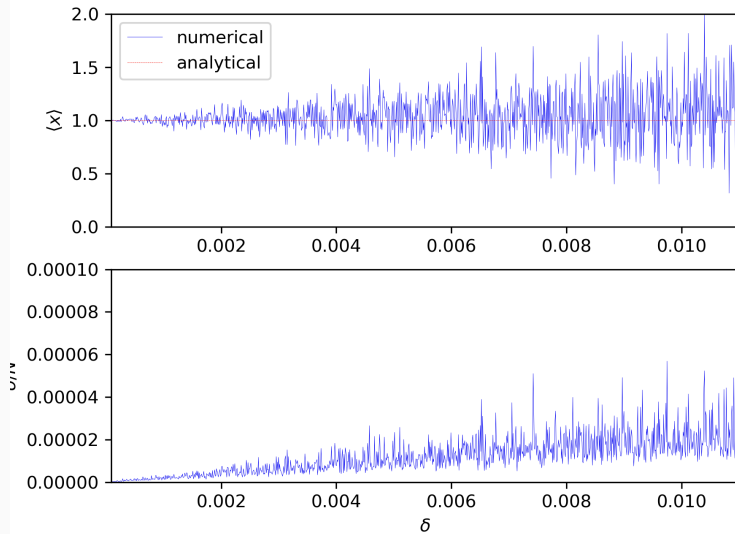
$$\langle x \rangle = \frac{\int_0^\infty x e^{-x} dx}{\int_0^\infty e^{-x} dx} = \frac{\overbrace{[-x e^{-x}]_0^\infty}^0 + \int_0^\infty e^{-x} dx}{\int_0^\infty e^{-x} dx} = 1 .$$

5.1 Method i

10 000 points are generated for $\delta = 0.0001, 0.0002, \dots, 0.1$ after the first 100 points are thrown. Starting parameter $x_0 = 1$ since this is x_i should converge to.

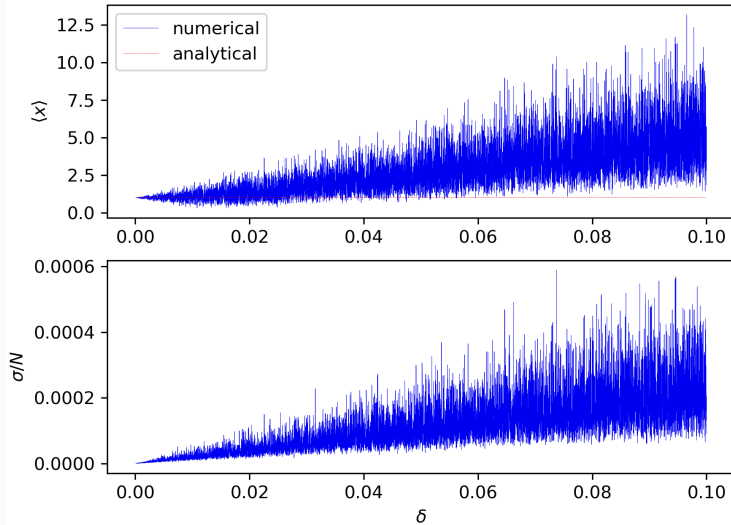
5.1 Results i

Calculating $\langle x \rangle = \frac{\int_0^\infty x e^{-x} dx}{\int_0^\infty e^{-x} dx}$ using the metropolis method.



5.1 Results ii

Calculating $\langle x \rangle = \frac{\int_0^\infty x e^{-x} dx}{\int_0^\infty e^{-x} dx}$ using the metropolis method.



5.1 Results iii

small δ : For small $\delta < 0.01$, $\langle x \rangle$ varies around the true value $\langle x \rangle = 1$.

large δ : For large $\delta > 0.01$, $\langle x \rangle$ increases linearly and fluctuates.

σ/N : σ/N increases as δ increases (which is expected). σ/N is several (≈ 5) orders of magnitudes smaller than the actual difference to the exact answer. Since $N = 10\,000$, $\sigma \sim$ the actual difference to the exact answer.

5.2

5.2 Project

Implement the Lennard-Jones potential with $\sigma = 1$ and $\varepsilon = 1$ in the template. Try to find a large MC step size that gives reasonable acceptance. How does the step size affect the convergence of the energy?

5.2 Background

Lennard-Jones potential.

The Lennard-Jones potential approximates the interaction between a pair of neutral atoms or molecules to the potential

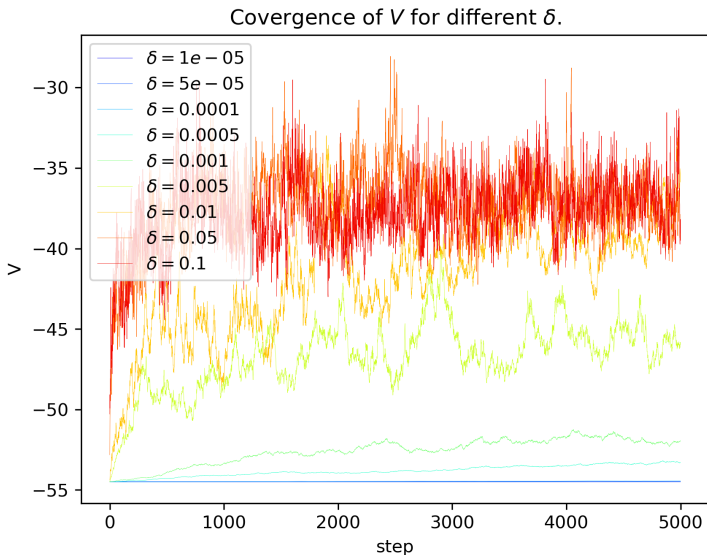
$$V_{LJ}(r) = 4\epsilon \left[\left(\frac{\sigma}{r} \right)^{12} - \left(\frac{\sigma}{r} \right)^6 \right] .$$

5.2 Method

A 5 000 step Monte-Carlo simulation is executed for each δ in

```
deltaArray = [0.00001, 0.00005, 0.0001, 0.0005,  
              0.001, 0.005, 0.01, 0.05, 0.1] .
```

5.2 Results i



5.2 Results ii

small δ : Poor convergence. Small variance.

large δ : Good convergence. Large variance.

best δ : $0.05 < \delta < 0.1$.

5.3

5.3 Project

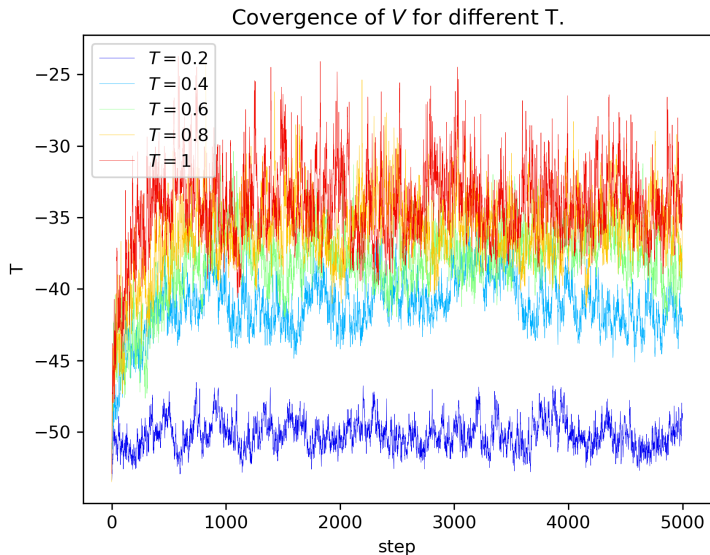
Run LJ simulations at $T = 1$ and 0.2. What differences in collective behavior do you observe between 1 and 0.2 at long times? A physics question: can you explain what you see?

5.3 Method

A 5 000 step Monte-Carlo simulation is executed with $\delta = 0.05$ for each T in

$$T_{\text{array}} = [0.2, 0.4, 0.6, 0.8, 1].$$

5.3 Results i



5.3 Results ii

convergence: Convergence is similar for all T .

small T : Small variance. Converges to small V .

large T : Large variance. Converges to large V .

physical interpretation: T is determined by the average kinetic energy of the particles. Large $T \implies$ higher movement \implies more chaotic system.
Furthermore, large $T \implies$ greater energy \implies greater V .

5.4

5.4 Project

Use MC to calculate the average energy and heat capacity at several temperatures between $T = 0.2$ and 1. Can you explain the behavior of the heat capacity by looking at the sampled configurations?

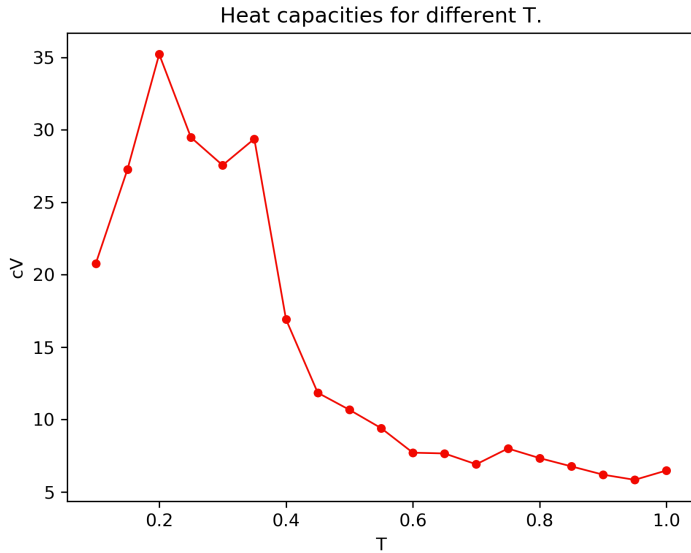
5.4 Method i

The heat capacity is determined for T in

$$TArray = [0.1, 0.15, 0.2, \dots, 1]$$

by averaging over $2\,000 \leq \text{step} \leq 10\,000$.

5.4 Results i



5.4 Results ii

The heat capacity decreases as the temperature increases.

Appendix

Available at:

<https://github.com/axelstr/SI1336-Simulations-and-Modeling/tree/master/5%20Monte%20Carlo>