

# How to compute observables

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## Thermodynamics

- the mean energy  $\langle E \rangle_\beta$  at the temperature  $T = 1/\beta$  is obtained by calculating the integral.

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- In a simulation, after a period corresponding the relaxation time, the system evolves at equilibrium and the mean energy is obtained by taking the arithmetic average

$$\langle E \rangle = \frac{1}{N_c} \sum_{i=1}^{N_c} E_i \quad (2)$$

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- $C_v$  is given by  $k_b \beta^2 (\langle E^2 \rangle - \langle E \rangle^2)$

## Time driven algorithm



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## Event-driven algorithm

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$$\langle A \rangle = \frac{1}{T} \sum_i^{N_c} A_i t_i \quad (4)$$

where  $t_i$  is time interval between the  $i$ th and  $i + 1$ th event (collision).



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- Microscopic pair density

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- Sum rules

$$\int \overline{\rho(\mathbf{r})} d\mathbf{r} = N$$

$$\int \overline{\rho(\mathbf{r}, \mathbf{r}')} d\mathbf{r} d\mathbf{r}' = N(N-1)$$

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$$g(\mathbf{r}, \mathbf{r}') = g(\mathbf{r} - \mathbf{r}')$$

For an isotropic and homogeneous system,

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Using the change of variable  $r, r' \rightarrow r, s = r' - r$  and integrating over  $r$

$$\rho g(s) = \frac{1}{N} \left\langle \sum_{i=1}^N \sum_{j=1, j \neq i}^N \delta(s - r_i + r_j) \right\rangle$$

## Implementation in a simulation.

- In a simulation box in  $3D$ , space discretization of  $g(r)$

$$\rho g(\Delta r(j + 0.5)) = \frac{1}{N(V_{j+1} - V_j)} 2N_p(r),$$

where  $N_p(r)$  is the number of distinct pairs whose center to center distances are between  $j\Delta r$  and  $(j + 1)\Delta r$  and where  $V_j = \frac{4\pi}{3}((j + 1)\Delta r)^3$ .

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- The maximum distance of  $g(r)$  is the half of the linear size of the simulation box.

- Definition

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$$S(k) = 1 + \frac{1}{N} \left\langle \int \int \exp(-ik(r - r')) \sum_{i=1}^N \sum_{j=1, i \neq j}^N \delta(r - r_i) \delta(r - r_j) dr dr' \right\rangle$$

which gives

$$S(k) = 1 + \frac{1}{N} \int \int \exp(-ik(r - r')) \rho(r, r') dr dr'.$$

- For an isotropic and uniform fluid

$$S(k) = 1 + \frac{\rho^2}{N} \int \int \exp(-ik(r - r'))g(r, r')drdr'$$

The radial distribution function only depends on  $|r - r'|$ . One then obtains

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- The Fourier transform only depends on the modulus  $|k|$

$$S(k) = 1 + 2\pi\rho \int r^2g(r) \int_0^\pi \exp(-ikr \cos(\theta)) \sin(\theta)d\theta dr$$

$$S(k) = 1 + 4\pi\rho \int_0^\infty r^2g(r) \frac{\sin(kr)}{kr} dr.$$

- For the Ising model, one defines the spin-spin time correlation function

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$$C(t) = \frac{1}{N} \sum_{i=1}^N \langle S_i(0) S_i(t) \rangle$$

- For a simple liquid made of point particles, the density autocorrelation function is

$$C(t) = \frac{1}{V} \int dr \langle \delta\rho(r, t) \delta\rho(r, 0) \rangle \quad (5)$$

where  $\delta\rho(r, t)$  denotes the local density fluctuation. This autocorrelation function

## Implementation

- Time average of a correlation function (or others quantities) at equilibrium uses a fundamental property of equilibrium systems, namely, time translational invariance
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$$C(t) = \frac{1}{NM} \sum_{j=1}^M \sum_{i=1}^N S_i(t_j)S_i(t_j + t).$$

- Define a timestep for  $C(t)$
- Define a matrix of  $N_c$  columns and  $N$  rows ( $N$  is the total number of spins) and once equilibrated, one defines a real vector of  $N_c$  components for  $C(t)$



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$$\tau_{eq} \ll N_c \Delta t_c \ll T_{sim} \quad (6)$$

## Implementation

- Consider a system at equilibrium described by the Hamiltonian  $\mathcal{H}_0$ . At time  $t = 0$ , an external force  $F(t)$  is applied and the additional Hamiltonian  $\mathcal{H}'$  is given by

$$\mathcal{H}' = -A(r^N)F(t)$$

where  $A(r^N)$  is the conjugate variable .

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- $F(t) \rightarrow 0$  when  $t \rightarrow \infty$
- The time evolution of the system is described by the Liouville equation

$$\begin{aligned}\frac{\partial f^{(N)}(r^N, p^N, t)}{\partial t} &= -i\mathcal{L}f^{(N)}(r^N, p^N, t) \\ &= \{\mathcal{H}_0 + \mathcal{H}', f^{(N)}(r^N, p^N, t)\} \\ &= -i\mathcal{L}_0 f^{(N)}(r^N, p^N, t) - \{A, f^{(N)}(r^N, p^N, t)\}F(t)\end{aligned}$$

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- Because the system was initially at equilibrium, one has

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$$f^{(N)}(r^N, p^N, 0) = C \exp(-\beta \mathcal{H}_0(r^N, p^N)),$$

- Since the external force is weak, one performs a perturbative expansion of  $f^{(N)}(r^N)$  around equilibrium. One writes

$$f^{(N)}(r^N, p^N, t) = f_0^{(N)}(r^N, p^N) + f_1^{(N)}(r^N, p^N, t)$$

- 

$$f_1^{(N)}(r^N, p^N, t) = - \int_{-\infty}^t \exp(-i(t-s)\mathcal{L}_0) \{A, f_0^{(N)}\} F(s) ds.$$

- Therefore, the variable  $\langle \Delta B(t) \rangle = \langle B(t) \rangle - \langle B(-\infty) \rangle$  evolves as

$$\langle \Delta B(t) \rangle = \int \int dr^N dp^N \left( f^{(N)}(r^N, p^N, t) - f_0^{(N)}(r^N, p^N) \right) B(r^N) dr^N dp^N.$$

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$$\begin{aligned} \langle \Delta B(t) \rangle &= - \int \int dr^N dp^N \int_{-\infty}^t \exp(-i(t-s)\mathcal{L}_0) \{A, f_0^{(N)}\} B(r^N) F(s) ds \\ &= - \int \int dr^N dp^N \int_{-\infty}^t \{A, f_0^{(N)}\} \exp(i(t-s)\mathcal{L}_0) B(r^N) F(s) ds \end{aligned}$$

- Calculating the Poisson bracket, one obtains that

$$\begin{aligned}\{A, f_0^{(N)}\} &= \sum_{i=1}^N \left( \frac{\partial A}{\partial r_i} \frac{\partial f_0^{(N)}}{\partial p_i} - \frac{\partial A}{\partial p_i} \frac{\partial f_0^{(N)}}{\partial r_i} \right) \\ &= -\beta \sum_{i=1}^N \left( \frac{\partial A}{\partial r_i} \frac{\partial \mathcal{H}_0^{(N)}}{\partial p_i} - \frac{\partial A}{\partial p_i} \frac{\partial \mathcal{H}_0^{(N)}}{\partial r_i} \right) f_0^{(N)} \\ &= -\beta i \mathcal{L}_0 A f_0^{(N)} \\ &= -\beta \frac{dA(0)}{dt} f_0^{(N)}\end{aligned}$$



# Linear response theory: results and transport coefficients

- Calculating the Poisson bracket, one obtains that

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Using

$$B(r^N(t)) = \exp(it\mathcal{L}_0)B(r^N(0)) \quad (7)$$

$$\langle \Delta B(t) \rangle = \beta \int_{-\infty}^t ds \left\langle \frac{dA(0)}{dt} B(t-s) \right\rangle F(s)$$

- $$\langle \Delta B(t) \rangle = \int_{-\infty}^{\infty} ds \chi(t, s) F(s) + \mathcal{O}(F^2)$$

where  $\chi(t, s)$  is the response function

# Linear response theory: results and transport coefficients

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- ① 
$$\chi(t) = \begin{cases} -\beta \frac{d}{dt} \langle A(0) B(t) \rangle & t > 0 \\ 0 & t < 0 \end{cases}$$

- ② A system cannot respond to a perturbation before it is applied. This property is a consequence of the causality.

$$\chi(t, s) = 0, \quad t - s \leq 0$$

- ③ The equilibrium response function is translationally invariant in time

$$\chi(t, s) = \chi(t - s) \tag{8}$$

- When  $A = B$ , The autocorrelation function is

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- By defining the linear integrated response as

$$R(t) = \int_0^t \chi(s) ds$$

The fluctuation-dissipation theorem is expressed as

$$R(t) = \begin{cases} \beta(C_A(0) - C_A(t)) & t > 0 \\ 0 & t < 0 \end{cases}$$

## Van Hove function



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## Van Hove function



$$\rho G(r, r'; t) = \langle \rho(r' + r, t) \rho(r', 0) \rangle$$

- This function can be expressed from the microscopic densities as

$$\rho G(r, r'; t) = \left\langle \sum_{i=1}^N \sum_{j=1}^N \delta(r' + r - r_i(t)) \delta(r - r_j(0)) \right\rangle$$

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- For a homogeneous system,  $G(\mathbf{r}, \mathbf{r}'; t)$  only depends on the relative distance. Integrating over volume, one obtains

$$G(\mathbf{r}, t) = \frac{1}{N} \left\langle \sum_{i=1}^N \sum_{j=1}^N \delta(\mathbf{r} - \mathbf{r}_i(t) + \mathbf{r}_j(0)) \right\rangle$$



## Van Hove function

- At  $t = 0$ , the Van Hove function  $G(\mathbf{r}, t)$  is simplified and one obtains

$$\begin{aligned} G(\mathbf{r}, 0) &= \frac{1}{N} \left\langle \sum_{i=1}^N \sum_{j=1}^N \delta(\mathbf{r} + \mathbf{r}_i(0) - \mathbf{r}_j(0)) \right\rangle \\ &= \delta(\mathbf{r}) + \rho g(\mathbf{r}) \end{aligned}$$

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- Function splitting

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- Physical interpretation

## Van Hove function

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- In the long time limit, the system loses memory of the initial configuration and the correlation functions become independent of the distance  $r$ .

$$\lim_{r \rightarrow \infty} G_s(r, t) = \lim_{t \rightarrow \infty} G_s(r, t) \simeq \frac{1}{V} \simeq 0$$

$$\lim_{r \rightarrow \infty} G_d(r, t) = \lim_{t \rightarrow \infty} G_s(r, t) \simeq \frac{N - 1}{V} \simeq \rho$$



## Intermediate scattering function



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- distinct part of the function

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- To be continued... 4-point correlation function, point-to-set function...