

**PHY 770 -- Statistical Mechanics**  
**11 AM-12:15 PM & 12:30-1:45 PM TR Olin 107**

Instructor: Natalie Holzwarth (Olin 300)  
 Course Webpage: <http://www.wfu.edu/~natalie/s14phy770>

**Lecture 7 & 8 -- Appendix A & Chapter 2**  
**Introduction to Probability and Its Role in Statistical Physics**

1. Probability distribution functions
2. Central limit theorem
3. Liouville theorem and its quantum equivalent
4. Relationship between entropy and notions from probability theory

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**PHY 770 Statistical Mechanics**

TR 12:30-1:45 PM OPL 107 <http://www.wfu.edu/~natalie/s14phy770/>

Instructor: Natalie Holzwarth Phone: 758-5510 Office: 300 OPL e-mail: natalie@wfu.edu

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**Course schedule for Spring 2014**

(Preliminary schedule -- subject to frequent adjustment.) Please note that makeup lectures (indicated in red) are scheduled for Tuesdays or Thursdays at 11 AM - 12:15 PM in Olin 107.

Lecture date	Text Reading	Topic	Assign.	Due date
1 Tue: 01/14/2014	Chap. 3	Review of macroscopic thermodynamics	as1	02/04/2014
2 Thu: 01/16/2014	Chap. 3	Review of macroscopic thermodynamics	as2	02/04/2014
3 Tue: 01/21/2014	Chap. 3	Thermodynamic potentials	as3	02/04/2014
4 Tue: 01/21/2014	Chap. 3	Thermodynamic stability	as4	02/04/2014
5 Thu: 01/23/2014	Chap. 3	Thermodynamic stability	as5	02/04/2014
Tue: 01/28/2014		IAWH out of town - no class		
Thu: 01/30/2014		IAWH out of town - no class		
6 Tue: 02/04/2014	Chap. 4	Phase transitions	as6	02/11/2014
7 Thu: 02/06/2014	Chap. 2	Microscopic analysis of entropy	as7	02/11/2014
8 Thu: 02/06/2014	Chap. 2	Microscopic analysis of entropy	as8	02/11/2014

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Some ideas from probability theory

Notation --

Random variable:  $X$

Possible value of  $X$ :  $x$

Probability of outcome  $x$ :  $P_X(x)$

Properties of probability function

Discrete case;  $x \Rightarrow x_i \quad i = 1, 2, \dots, N$

$P_X(x_i) \geq 0$

$\sum_{i=1}^N P_X(x_i) = 1$

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**Some ideas from probability theory -- continued**

Average value:  $\langle X \rangle \equiv \sum_{i=1}^N x_i P_X(x_i)$

Moment value:  $\langle X^n \rangle \equiv \sum_{i=1}^N x_i^n P_X(x_i)$

Standard deviation:  $\sigma_X = \sqrt{\langle X^2 \rangle - \langle X \rangle^2}$

For a continuous variable  $x$  where  $-\infty \leq x \leq \infty$ :

$P_X(x) \geq 0$

$\int_{-\infty}^{\infty} P_X(x) dx = 1$

$\langle X^n \rangle = \int_{-\infty}^{\infty} x^n P_X(x) dx$

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**Some ideas from probability theory -- continued**

Clever use of Fourier transforms; the characteristic function:

$\phi_X(k) \equiv \langle e^{ikx} \rangle = \int_{-\infty}^{\infty} e^{ikx} P_X(x) dx = \sum_{n=1}^{\infty} \frac{(ik)^n \langle X^n \rangle}{n!}$

Note that, using the inverse Fourier transform:

$P_X(x) = \frac{1}{2\pi} \int_{-\infty}^{\infty} e^{-ikx} \phi_X(k) dk$

Other useful relationships:

$\langle X^n \rangle = \frac{1}{i^n} \lim_{k \rightarrow 0} \frac{d^n \phi_X(k)}{dk^n}$

Cummulants:  $C_n(X)$

$\phi_X(k) = \exp\left(\sum_{n=1}^{\infty} \frac{(ik)^n}{n!} C_n(X)\right)$

$C_1(X) = \langle X \rangle$        $C_2(X) = \langle X^2 \rangle - \langle X \rangle^2$

$C_3(X) = \langle X^3 \rangle - 3\langle X^2 \rangle \langle X \rangle + 2\langle X \rangle^3$

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**Some ideas from probability theory -- continued**

Example:

$P_X(x) = \begin{cases} \frac{2}{\pi} \sqrt{1-x^2} & \text{for } |x| \leq 1 \\ 0 & \text{otherwise} \end{cases}$

$\phi_X(k) \equiv \langle e^{ikx} \rangle = \int_{-\infty}^{\infty} e^{ikx} P_X(x) dx = \frac{2}{\pi} \int_{-1}^1 e^{ikx} \sqrt{1-x^2} dx = \frac{2}{k} J_1(k)$

Using the ascending series expansion:

$\phi_X(k) = \frac{2}{k} J_1(k) = 1 - \frac{1}{4} \frac{k^2}{2!} + \frac{1}{8} \frac{k^4}{4!}$

From:  $\phi_X(k) \equiv \langle e^{ikx} \rangle = \int_{-\infty}^{\infty} e^{ikx} P_X(x) dx = \sum_{n=1}^{\infty} \frac{(ik)^n \langle X^n \rangle}{n!}$

$\langle X \rangle = \langle X^3 \rangle = \langle X^5 \rangle = \dots = 0$

$\langle X^2 \rangle = \frac{1}{4}$        $\langle X^4 \rangle = \frac{1}{8}$

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Some ideas from probability theory – continued

Example: Consider a random walk in one dimension for which the walker at each step is equally likely to take a step with displacement anywhere in the interval  $d-a \leq x \leq d+a$  ( $a < d$ ).

Each step is independent of the others. After  $N$  steps, the displacement of the walker is  $S = X_1 + X_2 + \dots + X_N$

What is the average  $\langle S \rangle$  and standard deviation  $\sigma_S$ ?

$$P_x(x) = \begin{cases} \frac{1}{2a} & \text{for } d-a \leq x \leq d+a \\ 0 & \text{otherwise} \end{cases}$$

For a single step :  $\phi_x(k) \equiv \langle e^{ikx} \rangle = \int_{-\infty}^{\infty} e^{ikx} P_x(x) dx = \frac{1}{2a} \int_{d-a}^{d+a} e^{ikx} dx = e^{ikd} \frac{\sin(ka)}{ka}$

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Some ideas from probability theory – continued

Characteristic function for a single step :

$$\phi_x(k) \equiv \langle e^{ikx} \rangle = e^{ikd} \frac{\sin(ka)}{ka}$$

Characteristic function for  $S$  ( $N$  steps) :

$$\phi_S(k) = (\phi_x(k))^N = \left( e^{ikd} \frac{\sin(ka)}{ka} \right)^N$$

Expansion in powers of  $k$  :

$$\phi_S(k) = \left( e^{ikd} \frac{\sin(ka)}{ka} \right)^N = 1 + (iNd)k - \left( \frac{1}{6} Na^2 + \frac{1}{2} N^2 d^2 \right) k^2 + \dots$$

$$\Rightarrow \langle S \rangle = Nd \quad \langle S^2 \rangle = \frac{1}{3} Na^2 + N^2 d^2$$

$$\sigma_S = \sqrt{\langle S^2 \rangle - \langle S \rangle^2} = \sqrt{\frac{N}{3}} a$$

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Some ideas from probability theory – continued

Typical probability functions

- Binomial distribution
- Gaussian distribution
- Poisson distribution

Binomial distribution

Consider a process with 2 outcomes:

- 0 with probability  $p$
- 1 with probability  $q=1-p$

For  $N$  “trials” of the process,  $n_0$  denotes the number outcomes 0 and  $n_1$  denotes the number of outcomes 1, with  $N=n_0+n_1$ .

$$P_N(n_1) = \frac{N!}{n_1!(N-n_1)!} p^{n_1} q^{(N-n_1)}$$

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**Binomial distribution continued:**

$$P_N(n_i) = \frac{N!}{n_i!(N-n_i)!} p^{n_i} q^{(N-n_i)}$$

Note that :

$$\sum_{n_i=0}^N P_N(n_i) = \sum_{n_i=0}^N \frac{N!}{n_i!(N-n_i)!} p^{n_i} q^{(N-n_i)} = (p+q)^N = 1$$

Can show that :

$$\langle n_i \rangle = Np$$

$$\langle n_i^2 \rangle = (Np)^2 + Npq$$

$$\sigma_N = \sqrt{Npq}$$

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**Example: Dice throws** 

On average, how many times must a die be thrown until "4" appears?

Let  $p$  = probability of getting 4 on one throw ( $p = 1/6$ )

Let  $q$  = probability of not getting 4 on one throw

Probability of first getting 4 on  $n$ th throw :  $P_n = pq^{n-1}$

Mean # of throws :  $m = \sum_{n=1}^{\infty} npq^{n-1} = p \frac{d}{dq} \sum_{n=0}^{\infty} q^n$

$$= p \frac{d}{dq} \frac{1}{1-q} = p \frac{1}{(1-q)^2} = \frac{1}{p}$$

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**Gaussian distribution**

Consider the binomial distribution in the limit of large  $N$  and large  $pN$ :

$$P_N(n) = \frac{N!}{n!(N-n)!} p^n q^{(N-n)}$$

$$\langle n \rangle = Np$$

Stirling approximation :  $n! \approx \sqrt{2\pi n} \left(\frac{n}{e}\right)^n$

$$P_N(n) \approx P_N(\langle n \rangle) \exp\left(-\frac{(n-\langle n \rangle)^2}{2Npq}\right) \rightarrow \frac{1}{\sqrt{2\pi}\sigma} \exp\left(-\frac{(n-\langle n \rangle)^2}{2\sigma^2}\right)$$

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Poisson distribution

Consider the binomial distribution in the limit of large  $N$  and  $pN = a \ll N$ :

$$P_N(n) = \frac{N!}{n!(N-n)!} p^n q^{(N-n)}$$

$$\langle n \rangle = Np = a$$

$$P_N(n) \approx \frac{a^n e^{-a}}{n!}$$

Note that :  $\sum_{n=0}^{\infty} \frac{a^n e^{-a}}{n!} = e^a e^{-a} = 1$

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Poisson distribution example

Consider a monolayer thin sheet of gold foil as a target for neutron scattering. Assume that the probability that in any given pulse of the beam the probability that the beam will scatter from the gold nuclei is given by the Poisson distribution with  $a=2$ . Determine the probability that  $n=0$  and that  $n=2$ .

$$P_{\text{Poisson}}(n) = \frac{a^n e^{-a}}{n!}$$

$$P_{\text{Poisson}}(0) = 2^0 e^{-2} = 0.135$$

$$P_{\text{Poisson}}(1) = \frac{2^1 e^{-2}}{1} = 0.271$$

$$P_{\text{Poisson}}(2) = \frac{2^2 e^{-2}}{2} = 0.271$$

$$P_{\text{Poisson}}(3) = \frac{2^3 e^{-2}}{3!} = 0.180$$

$$P_{\text{Poisson}}(4) = \frac{2^4 e^{-2}}{4!} = 0.090$$

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Central limit theorem

Consider  $N$  independent stochastic variables  $X_i, i=1,2,..N$ . What is the distribution of their sum

$$Y_N = (X_1 + \dots + X_N) / N$$

Characteristic function for  $Y_N$  :

$$\phi_Y(k) = \int_{-\infty}^{\infty} dx_1 \dots \int_{-\infty}^{\infty} dx_N e^{ik(x_1 + x_2 + \dots + x_N) / N} P_{X_1}(x_1) \dots P_{X_N}(x_N)$$

$$= (\phi_X(k/N))^N$$

$$\approx \left( 1 - \frac{1}{2} \frac{k^2 \sigma_X^2}{N^2} + \dots \right)^N \underset{N \rightarrow \infty}{\approx} \exp\left( -\frac{k^2 \sigma_X^2}{2N} \right)$$

$$P_Y(y) \approx \frac{1}{2\pi} \int_{-\infty}^{\infty} dk e^{-iky} \exp\left( -\frac{k^2 \sigma_X^2}{2N} \right) = \sqrt{\frac{N}{2\pi \sigma_X^2}} \exp\left( -\frac{Ny^2}{2\sigma_X^2} \right)$$

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**Central limit theorem**  
 Consider  $N$  independent stochastic variables  $X_i, i=1,2,\dots,N$ .  
 What is the distribution of their sum  
 $Y_N=(X_1+\dots+X_N)/N$

$$P_Y(y) = \sqrt{\frac{1}{2\pi\sigma_x^2/N}} \exp\left(-\frac{y^2}{2\sigma_x^2/N}\right)$$

→ Distribution function for  $Y$  is a Gaussian distribution centered at  $\langle x \rangle$  and with variance  $\sigma_x / \sqrt{N}$

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**Justification of statistical treatment of macroscopic systems**  
 Classical mechanics argument and the Liouville theorem

Liouville's theorem:  
 Imagine a collection of particles obeying the Canonical equations of motion in phase space.

Let  $D$  denote the "distribution" of particles in phase space:  
 $D = D(\{q_1 \dots q_{3N}\}, \{p_1 \dots p_{3N}\}, t)$   
 Liouville's theorem shows that:  
 $\frac{dD}{dt} = 0 \Rightarrow D$  is constant in time

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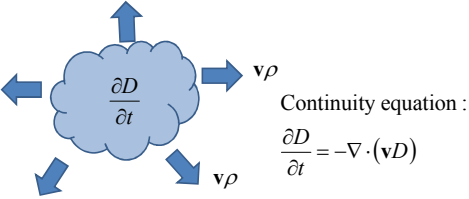
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**Proof of Liouville's theorem:**



Continuity equation:  
 $\frac{\partial D}{\partial t} = -\nabla \cdot (\mathbf{v}D)$

Note: in this case, the velocity is the  $6N$  dimensional vector:  
 $\mathbf{v} = (\dot{\mathbf{r}}_1, \dot{\mathbf{r}}_2, \dots, \dot{\mathbf{r}}_N, \dot{\mathbf{p}}_1, \dot{\mathbf{p}}_2, \dots, \dot{\mathbf{p}}_N)$   
 We also have a  $6N$  dimensional gradient:  
 $\nabla = (\nabla_{\mathbf{r}_1}, \nabla_{\mathbf{r}_2}, \dots, \nabla_{\mathbf{r}_N}, \nabla_{\mathbf{p}_1}, \nabla_{\mathbf{p}_2}, \dots, \nabla_{\mathbf{p}_N})$

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$$\frac{\partial D}{\partial t} = -\nabla \cdot (\mathbf{v}D)$$

$$= -\sum_{j=1}^{3N} \left[ \frac{\partial}{\partial q_j} (\dot{q}_j D) + \frac{\partial}{\partial p_j} (\dot{p}_j D) \right]$$

$$= -\sum_{j=1}^{3N} \left[ \frac{\partial D}{\partial q_j} \dot{q}_j + \frac{\partial D}{\partial p_j} \dot{p}_j \right] - D \sum_{j=1}^{3N} \left[ \frac{\partial \dot{q}_j}{\partial q_j} + \frac{\partial \dot{p}_j}{\partial p_j} \right]$$

$$\frac{\partial \dot{q}_j}{\partial q_j} + \frac{\partial \dot{p}_j}{\partial p_j} = \frac{\partial^2 H}{\partial q_j \partial p_j} + \left( -\frac{\partial^2 H}{\partial p_j \partial q_j} \right) = 0$$

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$$\frac{\partial D}{\partial t} = -\sum_{j=1}^{3N} \left[ \frac{\partial D}{\partial q_j} \dot{q}_j + \frac{\partial D}{\partial p_j} \dot{p}_j \right] - D \sum_{j=1}^{3N} \left[ \frac{\partial \dot{q}_j}{\partial q_j} + \frac{\partial \dot{p}_j}{\partial p_j} \right] = 0$$

$$\frac{\partial D}{\partial t} = -\sum_{j=1}^{3N} \left[ \frac{\partial D}{\partial q_j} \dot{q}_j + \frac{\partial D}{\partial p_j} \dot{p}_j \right]$$

$$\Rightarrow \frac{\partial D}{\partial t} + \sum_{j=1}^{3N} \left[ \frac{\partial D}{\partial q_j} \dot{q}_j + \frac{\partial D}{\partial p_j} \dot{p}_j \right] = \frac{dD}{dt} = 0$$

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**Complexity and entropy**

**Microscopic definition of entropy**

$$S(N, E, n) = k_B \ln(\mathcal{N}_N(E, n))$$

In this case, we have  $N$  particles having a total energy  $E$  and a macroscopic parameter  $n$ .

$\mathcal{N}_N(E, n)$  denotes the multiplicity of microscopic states having the same parameters. Each of these states are assumed to equally likely to occur.

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Example:  
 Suppose you have  $N$  spin-1/2 particles. How many microscopic states does the system have?

For  $N=10$ :  $\uparrow\downarrow\uparrow\uparrow\downarrow\downarrow\downarrow\downarrow\downarrow\uparrow$  total =  $2^{10}=1024$   
 For  $N=100$  total =  $2^{100}=10^{30}$

Now, consider  $N$  spin-1/2 particles with  $n \uparrow$ .

$$\mathcal{N}_N(n) = \frac{N!}{n!(N-n)!}$$

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Spin-1/2 system continued

Recall the binomial distribution : for fixed  $a$  and  $b$

$$(a+b)^N = \sum_{n=0}^N \frac{N!}{n!(N-n)!} a^{N-n} b^n$$

$$\Rightarrow \sum_{n=0}^N \frac{N!}{n!(N-n)!} = (1+1)^N = 2^N$$

Fraction of microscopic states for this system?

$$\mathcal{F}_N(n) = \frac{1}{2^N} \mathcal{N}_N(n) = \frac{1}{2^N} \frac{N!}{n!(N-n)!} = \frac{N!}{n!(N-n)!} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{N-n}$$

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Spin-1/2 system continued

Fraction of microscopic states for this system?

$$\mathcal{F}_N(n) = \frac{1}{2^N} \mathcal{N}_N(n) = \frac{1}{2^N} \frac{N!}{n!(N-n)!} = \frac{N!}{n!(N-n)!} \left(\frac{1}{2}\right)^n \left(\frac{1}{2}\right)^{N-n}$$

For  $N \rightarrow \infty$

$$\mathcal{F}_N(n) \approx \frac{1}{\sqrt{2\pi\sigma_N}} \exp\left(-\frac{(n-\langle n \rangle)^2}{2\sigma_N^2}\right)$$

where  $\langle n \rangle = N/2$  and  $\sigma_N = \sqrt{N}/2$

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Spin-1/2 system continued

Entropy for this system :

$$\mathcal{N}_N(n) = \frac{N!}{n!(N-n)!}$$

$$S(N, n) = k_B \ln(\mathcal{N}_N(n)) = k_B \ln\left(\frac{N!}{n!(N-n)!}\right)$$

Stirling approximation :  $N! \approx \sqrt{2\pi N} N^N e^{-N}$   
 $\ln(N!) \approx N \ln(N) - N$

$$S(N, n) = k_B \ln(\mathcal{N}_N(n)) \approx k_B \ln\left(\frac{N^N}{n^n (N-n)^{N-n}}\right)$$

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Spin-1/2 system continued

$$S(N, n) = k_B \ln(\mathcal{N}_N(n)) = k_B \ln\left(\frac{N!}{n!(N-n)!}\right)$$

$$S(N, n) = k_B \ln(\mathcal{N}_N(n)) \approx k_B \ln\left(\frac{N^N}{n^n (N-n)^{N-n}}\right)$$

$$S(N, \langle n \rangle) \approx k_B \ln\left(\frac{N^N}{\langle n \rangle^{\langle n \rangle} (N - \langle n \rangle)^{N - \langle n \rangle}}\right) = N k_B \ln(2)$$

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Relationship between probability function and entropy

Fraction of microscopic states for system

$$\mathcal{F}_N(n) = P_N(n) = \frac{1}{\mathcal{N}_N} \mathcal{N}_N(n) = \frac{1}{\mathcal{N}_N} \exp(S(N, n) / k_B)$$

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Spin-1/2 system continued – effects of Magnetic field

$H=0$        $H>0$

$$\langle E \rangle = -\mu \langle n \rangle H + \mu (N - \langle n \rangle) H = \mu N H - 2\mu \langle n \rangle H$$

Note that:  $\langle n \rangle = \frac{N}{2} - \frac{\langle E \rangle}{2\mu H}$

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Spin-1/2 system continued – effects of Magnetic field -- continued

Approximate entropy for this case (fixed  $\langle E \rangle, H$ );  
Stirling approximation :

$$S(N, \langle E \rangle, H) \approx k_B N \ln N - k_B \left( \frac{N}{2} - \frac{\langle E \rangle}{2\mu H} \right) \ln \left( \frac{N}{2} - \frac{\langle E \rangle}{2\mu H} \right) - k_B \left( \frac{N}{2} + \frac{\langle E \rangle}{2\mu H} \right) \ln \left( \frac{N}{2} + \frac{\langle E \rangle}{2\mu H} \right)$$

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Spin-1/2 system continued – effects of Magnetic field -- continued

Big leap:  
Assume the microscopic entropy function IS the same as the macroscopic entropy found in classical thermodynamics

$$\left( \frac{\partial S}{\partial E} \right)_{H,N} = \frac{1}{T}$$

$$S(N, \langle E \rangle, H) \approx k_B N \ln N - k_B \left( \frac{N}{2} - \frac{\langle E \rangle}{2\mu H} \right) \ln \left( \frac{N}{2} - \frac{\langle E \rangle}{2\mu H} \right) - k_B \left( \frac{N}{2} + \frac{\langle E \rangle}{2\mu H} \right) \ln \left( \frac{N}{2} + \frac{\langle E \rangle}{2\mu H} \right)$$

$$\left( \frac{\partial S}{\partial E} \right)_{H,N} = \frac{1}{T} = \frac{k_B}{2\mu H} \ln \left( \frac{N - \frac{\langle E \rangle}{2\mu H}}{N + \frac{\langle E \rangle}{2\mu H}} \right)$$

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