3. Lattice vibration

1 Boltzman distribution

1.1 Outline of Boltzman distribution

<u>Boltzman distribution</u> is the probability distribution or probability measure that gives the probability that a system will be in a certain state as a function of the energy of that state and the temperature of the system. The distribution is expressed in the billows,

$$n_{i} = \frac{N}{\sum_{j} exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right)} exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right) \propto exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right)$$
(1.1)

where, n_i is the probability of the system being in state *i*, *N* is the fixed large number of particles, ε_i is the energy of that state, k_B is <u>Boltzman's constant (1.380×10⁻²³ J/K)</u>, and *T* is temperature. In this chapter, the background knowledge for the derivation of boltzman distribution is presented and then the derivation is performed.

1.2 Fundamental concept of statistical mechanins

Entropy S is defined follows as Boltzmann's entropy formula,

$$S \equiv k_B \ln W \tag{2.1}$$

where the principle of equal a priori probabilities holds and the state with the largest entropy S appears most probably when it has the largest number of configurations W.

And the Temperature T is defined by using entropy S and internal energy E as follows.

$$1/T = \left(\frac{\partial S}{\partial E}\right)_{other \ conditions} \tag{2.2}$$

The above equation means that the rate of entropy increasing with increasing energy.

1.3 Lagrange multiplier

Lagrange multiplier is a strategy for finding local maxima and minima of a function subject to equality constrains. The method can be summarized Lagrangian function as follows,

$$L(x,y) = f(x,y) - \lambda g(x,y)$$
(3.1)

In order to find a point (a, b) where the maximum or minimum of a function f(x, y) subducted to the equality constraint g(x, y) = 0. Where λ is called Lagrange multiplier. If at least one of $\frac{\partial g}{\partial x}$ and

 $\frac{\partial g}{\partial y}$ is not zero at point (a, b), then there exists λ and the following holds at point (a, b, λ).

$$\frac{\partial L(a,b)}{\partial x} = \frac{\partial L(a,b)}{\partial y} = \frac{\partial L(a,b)}{\partial \lambda} = 0$$
(3.2)

On the other hand, the point (a, b) is different from points where f(x, y) has maxima and minima without the constraint g(x, y) = 0, the following hold.

$$\frac{\partial f}{\partial x} \neq 0 \text{ and } \frac{\partial f}{\partial y} \neq 0$$
 (3.3)

$$\frac{\partial L}{\partial x} = 0, \frac{\partial L}{\partial y} = 0 \tag{3.4}$$

Therefore,

$$\frac{\partial f}{\partial x} = \lambda \frac{\partial g}{\partial x}, \frac{\partial f}{\partial y} = \lambda \frac{\partial g}{\partial y}$$
(3.5)

Thus,

$$\left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y}\right) = \lambda \left(\frac{\partial g}{\partial x}, \frac{\partial g}{\partial y}\right)$$
(3.6)

We can visualize contours of f given by f(x, y) = d for various values of d, and the contour of g given by g(x, y) = c. The equation (3.6) means that $f = d_1 g = 0$ corves are parallel in the x-y plane. When (x, y) moves along g = 0, f does not change at a minimum/maximum (a,b). In other words, $f = d_1$ and g = 0 are parallel at (a,b) (fig.1). The ratio of the change of f(x, y) to the change of g(x, y)by changing parameters (x, y), where f(x, y) and g(x, y) are not constant.

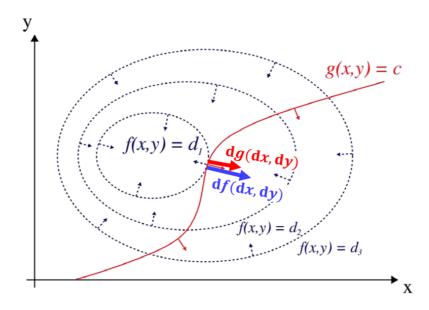


Fig. 1. The red curve shows the constraint g(x, y) = c. The blue curves are contours of f(x, y). The point where the red constraint tangentially touches a blue contour is the maximum of f(x, y) along the constraint, since $d_1 > d_2$.

1.4 Derivation of Boltzmann distribution

Considering a system composed of the fixed number of N particles with a fixed total energy E, energy of a particle is ε_i , and the number of particles having an energy ε_i is n_i . Then the total number of particles N and the total energy of the system E can be expressed as follows.

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$$\mathbf{V} = \sum_{i=0}^{\infty} n_i \tag{4.1}$$

$$E = \sum_{i=0}^{\infty} n_i \varepsilon_i \tag{4.2}$$

And here, the number of configuration of the system $(n_1, n_2, n_3, ...)$, *W* and the entropy of the system are as billows.

$$W = \frac{N!}{n_0! \, n_1! \, n_2! \dots} \tag{4.3}$$

$$S = k_B \ln W = k_B \ln \frac{N!}{n_0! n_1! n_2! \dots}$$
(4.4)

$$= k_B \left(\ln N! - \sum_{i=0} n_i! \right)$$

Here, the Stirling's approximation is as follows.

$$\ln N! \cong N \ln N - N \tag{4.5}$$

Using the Stirling's approximation (4,5), $\ln W$ in (4.4) becomes as billows,

$$\ln W \cong N \ln N - N - \sum_{i=0}^{\infty} n_i \ln n_i - n_i$$

= $N \ln N - \sum_{i=0}^{\infty} n_i \ln n_i - n_i$
- $[N - \sum_{i=0}^{\infty} n_i]$
= $N \ln N - \sum_{i=0}^{\infty} n_i \ln n_i$ (4.6)

Therefore, the conditions for the largest $\ln W$ will be $d \ln W = 0$. And the total number of particles and the total energy of the system are fixed as billow from (4.1) and (4.2),

$$\mathrm{d}N = \sum_{i=0}^{\infty} \mathrm{d}n_i = 0 \tag{4.7}$$

$$dE = \sum_{i=0}^{\infty} n_i d\varepsilon_i = 0 \tag{4.8}$$

Using (4.6), the change in the logarithmic number of microstates, $\ln W$ is as follows,

$$d \ln W = d(N \ln N - \sum_{i=0}^{\infty} n_i \ln n_i)$$

= $(dN + Nd \ln N)$
 $- \sum_{i=0}^{\infty} (dn_i \ln n_i + n_i d \ln n_i)$
= $-\sum_{i=0}^{\infty} (dn_i \ln n_i + \frac{n_i dn_i}{n_i})$
= $-\sum_{i=0}^{\infty} (1 + \ln n_i) dn_i$
 $\cong -\sum_{i=0}^{\infty} \ln n_i dn_i$
(4.9)

Applying the method of Lagrange multiplier to obtain the maximum $\ln W$, which indicates the most probable state, under conditions of fixed *N* and *E*.

$$L = \ln W + \alpha N - \beta E \tag{4.10}$$

Where, α and β are Lagrange multiplier constant. Differentiating both sides of (4.10), we obtain

$$dL = d \ln W + \alpha dN - \beta dE$$

= $-\sum_{i=0}^{\infty} \ln n_i dn_i - \alpha \sum_{i=0}^{\infty} dn_i$
 $-\beta \sum_{i=0}^{\infty} \varepsilon_i dn_i$ (4.11)

here, using (4.7) (4.8) (4.9), (4.11) becomes as billows,

$$dL = -\sum_{i=0}^{\infty} (\ln n_i + \alpha + \beta \varepsilon_i) dn_i$$
(4.12)

Therefore,

$$dL = \sum_{i=0}^{\infty} (\ln n_i + \alpha + \beta \varepsilon_i) dn_i = 0$$
(4.13)

Thus, we obtain the following.

 $\ln n_i + \alpha + \beta \varepsilon_i = 0 \tag{4.14}$

From (4.14), the Boltzmann distribution (3.1) becomes as follows,

$$n_{i} = \exp(-\alpha - \beta \varepsilon_{i}) = \exp(-\alpha) \exp(-\beta \varepsilon_{i})$$

= $A \exp(-\beta \varepsilon_{i})$ (4.15)

Where, A is $\exp(-\alpha)$. And n_i and ε_i are balanced to maximize W at constant N and E. From here, determine β from (4.14). Multiplying both sides of (4.14) by $\sum_i n_i$, we obtain,

$$\sum_{i} n_{i} \ln n_{i} + \alpha \sum_{i} n_{i} + \beta \sum_{i} n_{i} = 0$$
(4.16)

Using (4.6), (4.16) is changed as follow.

$$N\ln N - \ln W + \alpha \sum_{i} n_{i} + \beta \sum_{i=0}^{\infty} n_{i} \varepsilon_{i} = 0$$

$$(4.17)$$

By multiplying by k_{B} , (4.16) becomes as follow.

$$k_B N \ln N - k_B \ln W + k_B \alpha \sum_i n_i + k_B \beta \sum_{i=0}^{\infty} n_i \varepsilon_i$$

= 0 (4.18)

From the definition of entropy (3.2), (4.18) becomes as follow.

$$k_B N \ln N - S + \alpha k_B N + \beta k_B E = 0 \tag{4.19}$$

Thus,

$$S = k_B N \ln N + \alpha k_B N + \beta k_B E \tag{4.20}$$

Differentiating (4.20) by E,

$$\frac{\mathrm{d}S}{\mathrm{d}E} = \frac{\mathrm{d}}{\mathrm{d}E} (k_B N \ln N + \alpha k_B N + \beta k_B E) = \beta k_B \tag{4.21}$$

From the definition of the temperature, T, $\frac{dS}{dE} = \frac{1}{T}$

$$\beta k_B = \frac{1}{T} \tag{4.22}$$

Thus,

$$\beta = \frac{1}{k_B T} \tag{4.23}$$

From here, determine the factor A of (4.15). By substituting (4.23) into (4.14),

$$\ln n_{\rm i} + \alpha + \frac{\varepsilon_i}{k_B T} = 0 \tag{4.24}$$

$$n_{\rm i} = \exp(-\alpha)\exp(-\frac{\varepsilon_i}{k_B T}) = A\exp(-\frac{\varepsilon_i}{k_B T})$$
(4.25)

By substituting (4.25) into (4.1),

$$N = \sum_{i} A \exp(-\frac{\varepsilon_{i}}{k_{B}T})$$
(4.26)

Thus, A is as billows.

$$A = \frac{N}{\sum_{i} \exp(-\frac{\varepsilon_{i}}{k_{B}T})}$$
(4.27)

Thus, the Boltzmann distribution can be expressed as follows (3.1).

$$n_{i} = \frac{N}{\sum_{j} exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right)} exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right) \propto exp\left(-\frac{\varepsilon_{i}}{k_{B}T}\right)$$
(1.1)