



Ensembles I

Counting microstates

Chris Oostenbrink
 Institute of molecular modeling and simulation
 University of Natural Resources and Life Sciences



Institute of molecular modeling and simulation

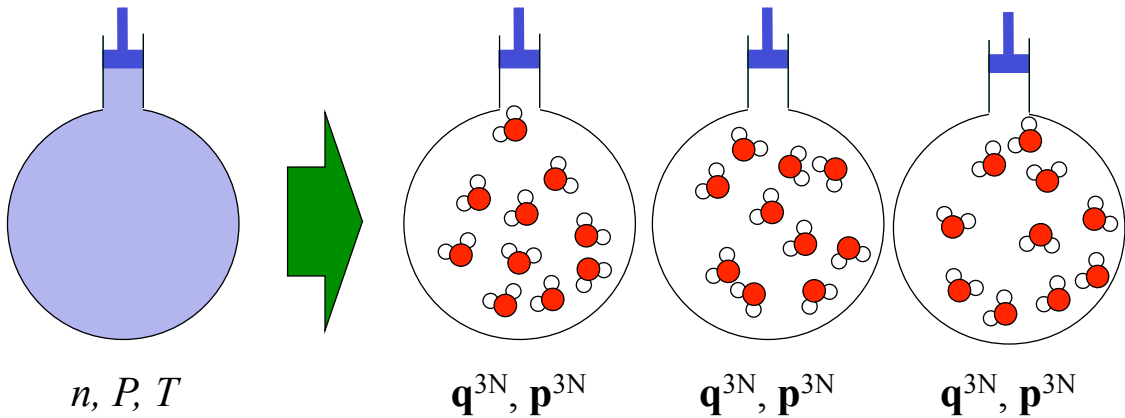


Schedule

Time	Mon. 20/11/17	Tue. 21/11/17	Wed. 22/11/17	Thur. 23/11/17	Fri. 24/11/17	Mon. 27/11/17	Tue. 28/11/17	Wed. 29/11/17	Thur. 30/11/17	Fri. 01/12/17
9:00	Free Time	Welcome <small>JAG</small>	Free Time	Free Time	Free Time	Free Time	Free Time	Free Time	Free Time	Free Time
9:30		Lecture 1: Overview and Introduction <small>JAG</small>	Lecture 4: Thermodynamics <small>JAG</small>	Lecture 7: Classical Mechanics I <small>JAG</small>	Lecture 10: Ensembles I <small>CO</small>	Lecture 13: Free energies: reaction coordinates <small>JAG</small>	Lecture 16: Calculating properties from simulations <small>JAG</small>	Lecture 19: Electrostatics <small>WvG</small>	Lecture 22: On the ethics of the academic endeavour: where do we go? <small>WvG</small>	Lecture 24: Left-overs/ questions and future perspectives <small>WvG</small>
10:15		Break <small>JAG</small>	Break <small>JAG</small>	Break <small>JAG</small>	Break <small>CO</small>	Break <small>JAG</small>	Break <small>JAG</small>	Break <small>WvG</small>		Break <small>WvG</small>
10:30		Lecture 2: Molecular Simulations MD/SD/MC <small>CO</small>	Lecture 5: Force-Field Development <small>CO</small>	Lecture 8: Classical Mechanics II <small>JAG</small>	Lecture 11: Ensembles II <small>JAG</small>	Lecture 14: Boundary Conditions I <small>WvG</small>	Lecture 17: Comparison with Experiments <small>WvG</small>	Lecture 20: Polarization <small>WvG</small>		Lecture 25: Students plans I
11:15		Coffee Break <small>CO</small>	Coffee Break <small>CO</small>	Coffee Break <small>JAG</small>	Coffee Break <small>JAG</small>	Coffee Break <small>WvG</small>	Coffee Break <small>WvG</small>	Coffee Break <small>WvG</small>	Coffee Break <small>WvG</small>	Coffee Break
11:45		Lecture 3: How to simulate using GROMOS <small>CO</small>	Lecture 6: Structure Refinement <small>CO</small>	Lecture 9: Analyzing with GROMOS <small>CO</small>	Lecture 12: Free energies: alchemy <small>CO</small>	Lecture 15: Boundary Conditions II <small>WvG</small>	Lecture 18: Searching & Enhanced Sampling <small>WvG</small>	Lecture 21: Multi-resolution simulations <small>WvG</small>	Lecture 23: QM/MM <small>WvG</small>	Lecture 26: Students plans II
12:30		Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Break for lunch, self-study, discussion.*	Tutorial 10: **
14:00	Registration	Tutorial 2: Running MD GROMOS Tutorial	Tutorial 3: Running MD Students Plans	Tutorial 4: Running MD Students Plans	Tutorial 5: Running MD Students Plans	Tutorial 6: Statistical Mechanics exercises	Tutorial 7: Analyzing MD: GROMOS Tutorial	Tutorial 8: Analyzing MD: Students Plans	Tutorial 9: Analyzing MD: Students Plans	How to prepare a barbecue
16:00	Coffee Break	Coffee Break	Coffee Break	Coffee Break	Coffee Break	Coffee Break	Coffee Break	Coffee Break	Coffee Break	
17:30	End of session	End of session	End of session	End of session	End of session	End of session	End of session	End of session	End of session	
20:30	Free time	Free time	Free time	Free time	Beer, Science & Friendship Good Stock Bar	Free time	Free time	Free time	Free time	Farewell

What is statistical mechanics about?

- Provide the link between macroscopic properties and the molecular (nanoscopic) particles.



“Macroscopic” view
(e.g. experiment)

*Description of a system in
terms of measurable
(macroscopic)
thermodynamic quantities*

“Microscopic” view
(e.g. molecular simulation)

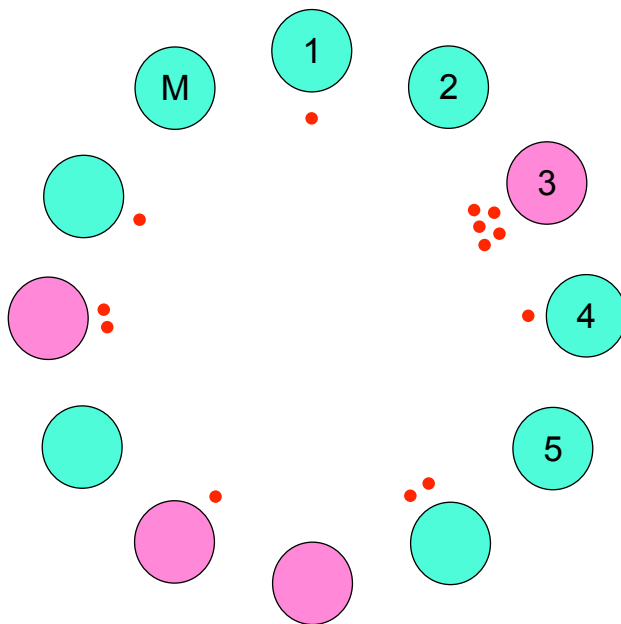
*Description of a system in
terms of atomic details*

Outline

- A game
- States and entropy
- Microcanonical ensemble

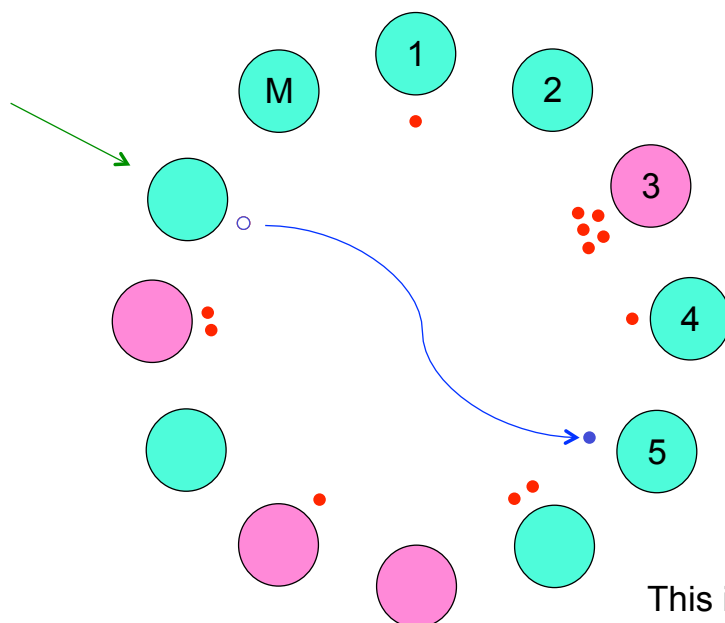
Balls and buckets

- A game with N children sitting in a circle, N_g girls, N_b boys
- Teacher divides M balls at random over them



Balls and buckets

- The **teacher** picks a random kid
- If it has a ball, it has to give it to another random kid



This is repeated until the kids are bored

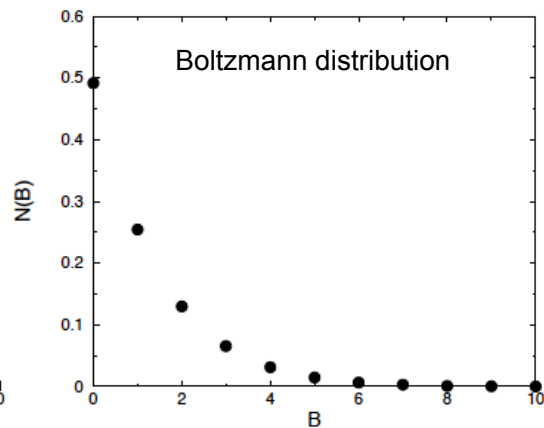
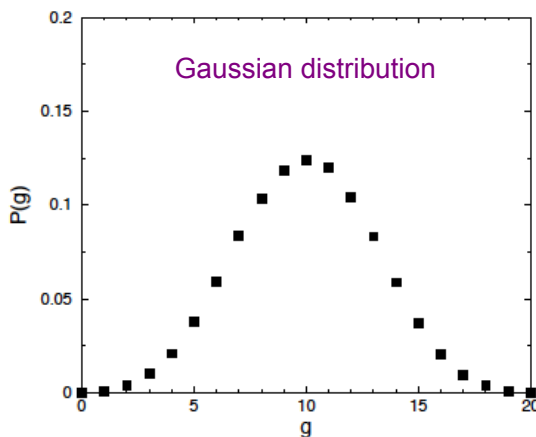
Balls and buckets

- The average number of balls per kid is $\langle B \rangle = M/N$
- The probability that any kid has 0, 1, 2, 3, ... balls appears to be

$$P(B) = C e^{-B/\langle B \rangle}$$

- What is the average number of balls collected by the girls?

$$\langle g \rangle = f_g M = \frac{N_g}{N} M \quad P(g) = C' e^{-(g - \langle g \rangle)^2 / 2\sigma^2}$$

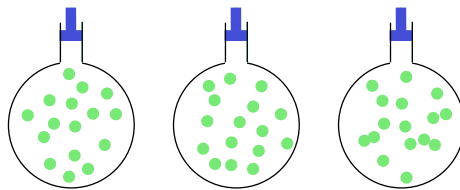


Balls and buckets

- What if the teacher was not fair initially?
 - The 'correct' distributions will appear after enough steps
- Analogies:
 - The number of balls remains constant (constant energy)
 - No kid can have less than 0 balls (ground-state energy is minimum)
- If M approaches infinity, we reach the thermodynamic limit
 - P(g) becomes very sharply peaked around $\langle g \rangle$
- Two approaches to get P(B) (ergodicity)
 - Follow one kid for a long time: time average (MD)
 - Monitor the averages over all kids: ensemble average (MC)

Isolated systems

- No heat transfer, no work (mechanical or chemical):
 - Constant number of particles, volume and energy: NVE



- At constant E , any individual particle can still have an energy E_k
- If N_k particles have energy E_k we can write that

$$E = \sum_k N_k E_k \quad N = \sum_k N_k$$

- For the game, this was the total number of kids N and the total number of balls M . With N_k kids having k balls:

$$M = \sum_k N_k k \quad N = \sum_k N_k$$

Isolated systems

- The individual particles have an energy E_k
- They may exchange energy through collisions
 - (passing the balls around)
- The distribution of positions and velocities does not depend on the kind of collisions
- Rather we make a **fundamental assumption**:
 - All possible configurations of the system with the same overall energy are equally likely



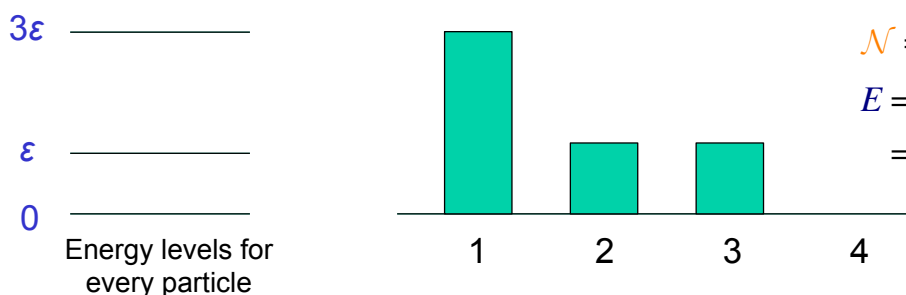
Let's look at an example

- Individual particles have three energy levels:
 - $E_0 = 0$ $E_1 = \epsilon$ $E_2 = 3\epsilon$
- For N particles we can define the system state S
 - $S = (k_1, k_2, \dots, k_N)$ (specify energy level for all particles)
- And we can write the system distribution
 - $\mathcal{N} = (N_0, N_1, N_2)$ (specify number of particles at every level)

$$E = \sum_k N_k E_k \qquad N = \sum_k N_k$$

Possible state of the system

- Example for $N = 4$



$$S = (2, 1, 1, 0)$$

$$\mathcal{N} = (1, 2, 1)$$

$$E = 1 \cdot 3\epsilon + 2 \cdot \epsilon + 1 \cdot 0 = 5\epsilon$$

Enforce total energy

- If the total energy is forced to be 3ϵ list all possible states

#	$S = (k_1, k_2, k_3, k_4)$				$\mathcal{N} = (N_0, N_1, N_2)$		
	k_1	k_2	k_3	k_4	N_0	N_1	N_2
1	0	0	0	2	3	0	1
2	0	0	2	0	3	0	1
3	0	2	0	0	3	0	1
4	2	0	0	0	3	0	1
5	1	1	1	0	1	3	0
6	1	1	0	1	1	3	0
7	1	0	1	1	1	3	0
8	0	1	1	1	1	3	0



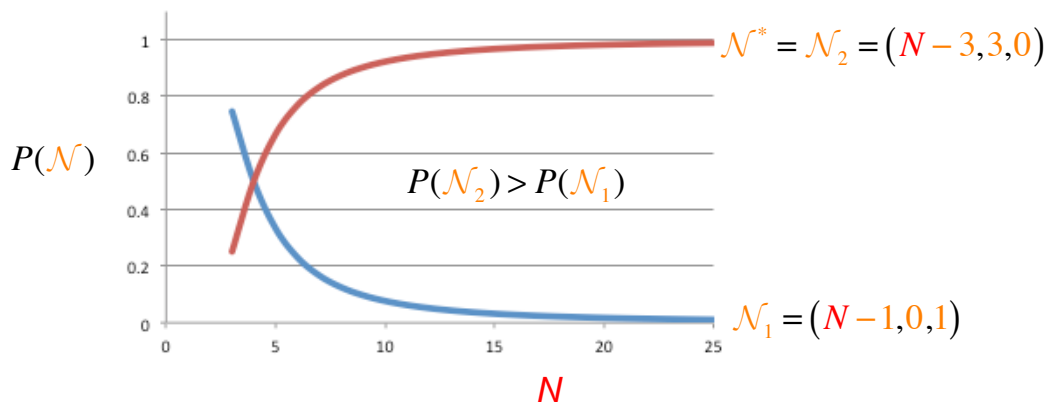
- Eight states are possible, with two distinct distributions
- The fundamental assumption says that all eight are equally likely

Larger values of N

- Let's increase the number of particles, but keep the energy at 3ϵ
- For $N = 5$ we have:
 - 5 possibilities to get $\mathcal{N}_1 = (4,0,1)$
 - 10 possibilities to get $\mathcal{N}_2 = (2,3,0)$

Every individual configuration is equally likely

- For large N , the number of possibilities for $\mathcal{N}_1 = (N-1,0,1)$ is much smaller than for $\mathcal{N}_2 = (N-3,3,0)$
- For large N , a single distribution becomes dominant: \mathcal{N}^*



The possibilities of distributing

- We need to calculate the number of possibilities to get a given distribution
- If I have N objects, 1, 2, 3, ... N , I can list them in $N!$ ways
 - $N=3$: (1 2 3) / (1 3 2) / (2 1 3) / (2 3 1) / (3 1 2) / (3 2 1)
 - First pick one (N possibilities), pick the next ($N-1$ possibilities), etc. until I have to pick the last one (1 possibility)
 - $N(N-1)(N-2)\dots 1 = N!$
- If I have N objects and I want to distribute them in 2 groups
 - $N=3, M=1$: (1 | 2 3) / (2 | 1 3) / (3 | 1 2)
 - We don't want to count (1 | 3 2) / (2 | 3 1) / (3 | 2 1) separately
 - Total number of possibilities is $3! / (1! 2!) = 6 / (1 \cdot 2) = 3$
 - In general, dividing N objects in two classes with M and $N-M$

$$W = \binom{N}{M} = \frac{N!}{M!(N-M)!}$$

Number of possibilities

- Generalizing

$$W = \binom{N}{M} = \frac{N!}{M!(N-M)!}$$

- from two states to an infinite amount of states, we get

$$W(\mathcal{N}) = \frac{N!}{\prod_k N_k!}$$

- This will be a big number, so we write

$$\ln W(\mathcal{N}) = \ln N! - \ln \prod_k N_k! = \ln N! - \sum_k \ln N_k!$$

- and now remember Stirling's rule for big N :

$$\ln N! \approx N \ln N - N$$

- and as N_k is also big

$$\ln N_k! \approx N_k \ln N_k - N_k$$

Number of possibilities

- So, then we can write

$$\begin{aligned} \ln W(\mathcal{N}) &= \ln N! - \sum_k \ln N_k! \\ &= N \ln N - N - \sum_k N_k \ln N_k + \sum_k N_k \\ &= N \ln N - \sum_k N_k \ln N_k \\ &= - \sum_k N_k (\ln N_k - \ln N) \\ &= -N \sum_k \frac{N_k}{N} \left(\ln \frac{N_k}{N} \right) \\ &= -N \sum_k P(k|\mathcal{N}) \ln P(k|\mathcal{N}) \end{aligned}$$

Redistribution of particles over the states at constant NVE is a spontaneous process, for which Entropy increases

Entropy is proportional to N (extensive variable)

Boltzmann's suggestion

- Boltzmann postulates that the statistical entropy is

$$S(\mathcal{N}) = k_B \ln W(\mathcal{N})$$

- In the thermodynamic limit $N \rightarrow \infty$
 - The statistical entropy becomes the thermodynamic entropy

$$S(U, V, N) = k_B \ln W(\mathcal{N}^*) = -Nk_B \sum_k P(k) \ln P(k)$$

- Similarly, the thermodynamic energy is approached by the total energy of the particles

$$U = E = \sum_k N_k E_k$$

Boltzmann

- You can visit him at Zentralfriedhof in Vienna

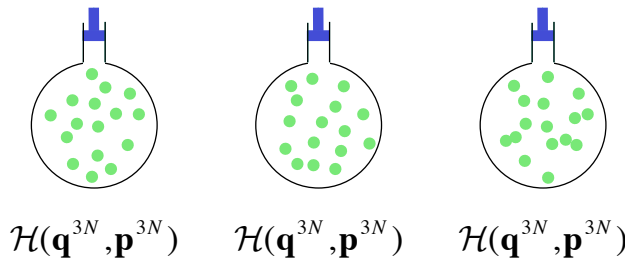
$$S(\mathcal{N}) = k_B \ln W(\mathcal{N})$$

- As we did during a workshop on Entropy in May 2014



Ensembles

- By now, we have implicitly defined the microcanonical ensemble
- At constant N, V, E , all configurations have the same probability
- If we have a system of N particles, it is completely defined by the positions \mathbf{q} and momenta \mathbf{p} of the individual particles
- We can calculate their energy through the Hamiltonian $\mathcal{H}(\mathbf{q}^{3N}, \mathbf{p}^{3N})$



... infinite number
of microstates

$$P_{NVE}(\mathbf{q}, \mathbf{p}) = \begin{cases} 1/W(E, V, N) & \text{if } E \leq \mathcal{H}(\mathbf{q}, \mathbf{p}) \leq E + \Delta \\ 0 & \text{otherwise} \end{cases}$$

Microcanonical ensemble

- Transform from a discrete enumeration to continuous variables
- The number of possible configurations is called the density of states or the partition function

$$W(E, V, N) = \Omega(E, V, N) = \frac{1}{N!} \frac{1}{h^{3N}} \iint d\mathbf{q} d\mathbf{p}$$

- h is the quantum mechanical volume of a microstate in phase space (has unit of mass-length²/time)
- The factor $N!$ comes from the indistinguishability of the particles
- The normalization holds

$$\frac{1}{N!} \frac{1}{h^{3N}} \iint P(\mathbf{p}, \mathbf{q}) d\mathbf{q} d\mathbf{p} = 1$$

- and we can calculate the expectation value (average) of any quantity A

$$\langle A \rangle_{NVE} = \frac{1}{N!} \frac{1}{h^{3N}} \iint P(\mathbf{p}, \mathbf{q}) A(\mathbf{p}, \mathbf{q}) d\mathbf{q} d\mathbf{p}$$

Let's go back to a discrete example

- A system of N molecules
- Every molecule can be in one of three energy states
 - $E_0 = 0$ $E_1 = \varepsilon$ $E_2 = 2\varepsilon$
 - N_k denotes the numbers in state k

$$N_0 + N_1 + N_2 = N$$

$$N_1\varepsilon + 2N_2\varepsilon = U$$

- From this, we can write

$$N_2 = \frac{U - N_1\varepsilon}{2\varepsilon} \quad N_0 + N_1 + \frac{U - N_1\varepsilon}{2\varepsilon} = N$$

$$\left(1 - \frac{1}{2}\right)N_1 = N - N_0 - \frac{U}{2\varepsilon} \Rightarrow N_1 = 2N - 2N_0 - \frac{U}{\varepsilon}$$

$$N_1 = \frac{U - 2N_2\varepsilon}{\varepsilon} \quad N_0 + \frac{U - 2N_2\varepsilon}{\varepsilon} + N_2 = N$$

$$(-2 + 1)N_2 = N - N_0 - \frac{U}{\varepsilon} \Rightarrow N_2 = N_0 - N + \frac{U}{\varepsilon}$$

By applying the constraints of constant N and U , we can write N_1 and N_2 as function of N_0 , N and U

Another example

- The distribution is then

$$\begin{aligned} \mathcal{N} &= (N_0, N_1, N_2) \\ &= (N_0, 2N - 2N_0 - U/\varepsilon, N_0 - N + U/\varepsilon) \end{aligned}$$

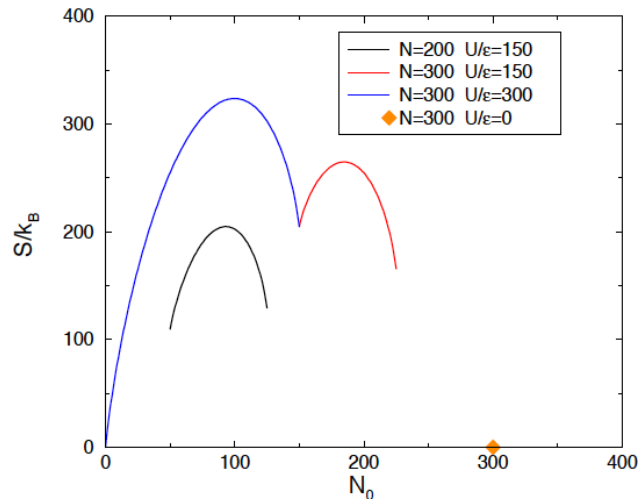
- And we can also write

$$W(\mathcal{N}) = \frac{N!}{N_0!N_1!N_2!}$$

$$S = k_B \ln W(\mathcal{N}) = k_B \ln \left[\frac{N!}{N_0!(2N - 2N_0 - U/\varepsilon)!(N_0 - N + U/\varepsilon)!} \right]$$

- this is something we can even calculate in Excel (for $N = 50$)

Another example



- For $N = 300$ and $U = 0$, there is only one point, because there is only one possible distribution: $\mathcal{N} = (300, 0, 0)$ and $S = 0$
- For $N = 300$ and $U/\epsilon = 300$, maximum entropy for $\mathcal{N}^* = (100, 100, 100)$
- For $N = 300$ and $U/\epsilon = 150$, max S at $N_0 = 180$; $P(0) = 180/300 = 0.6$
- For $N = 200$ and $U/\epsilon = 150$, max S at $N_0 = 80$; $P(0) = 80/200 = 0.4$
 - More particles, same energy: more have to be in the ground state

Determine the most likely distribution

- We want to find the distribution that maximizes

$$\ln W(\mathcal{N}) = - \sum_k N_k \ln \frac{N_k}{N}$$

- Under the conditions that N and E are constant

$$N = \sum_k N_k \quad E = \sum_k N_k E_k$$

- Formulate the constraints as Lagrange multipliers

$$N - \sum_k N_k = 0 \quad E - \sum_k N_k E_k = 0$$

- And define an alternative function to maximize

Determine the most likely distribution

- We want to maximize the function

$$\begin{aligned}\Phi(\mathcal{N}) &= \ln W(\mathcal{N}) - \alpha \left(N - \sum_k N_k \right) + \beta \left(E - \sum_k N_k E_k \right) \\ &= - \sum_k N_k \ln \frac{N_k}{N} - \alpha \left(N - \sum_k N_k \right) + \beta \left(E - \sum_k N_k E_k \right)\end{aligned}$$

- The derivative with respect to any N_k gives

$$\frac{\partial \Phi(\mathcal{N})}{\partial N_k} = - \ln \frac{N_k}{N} - N_k \frac{N}{N_k} \frac{1}{N} + \alpha - \beta E_k = 0$$

- and from that we derive

$$\begin{aligned}\ln \frac{N_k}{N} &= -1 + \alpha - \beta E_k \\ P(k) &= e^{\alpha-1} e^{-\beta E_k} = Q^{-1} e^{-\beta E_k}\end{aligned}$$

- So the probability of a state k is proportional to $\exp(-\beta E_k)$

What is Q?

- The normalization constant Q is readily determined

$$N = \sum_k N_k$$

- so that means that

$$\sum_k P(k) = \sum_k \frac{N_k}{N} = \frac{N}{N} = 1$$

- and thus

$$\begin{aligned}\sum_k P(k) &= \sum_k Q^{-1} e^{-\beta E_k} = 1 \\ &= \frac{1}{Q} \sum_k e^{-\beta E_k} = 1\end{aligned}$$

$$Q = \sum_k e^{-\beta E_k}$$

Q is the partition function of the canonical ensemble. We will see it come back often

and what is β ?

- That is not so straightforward. We start from the characteristic equation for the total energy

$$dU = -PdV + \sum_{i=1}^N \mu_i dn_i + TdS$$

- and we look at the temperature

$$T = \left(\frac{\partial U}{\partial S} \right)_{N,V}$$

- Using a chain rule towards β

$$T = \left(\frac{\partial U}{\partial \beta} \right)_{N,V} \left(\frac{\partial \beta}{\partial S} \right)_{N,V} = \left(\frac{\partial U}{\partial \beta} \right)_{N,V} / \left(\frac{\partial S}{\partial \beta} \right)_{N,V}$$

towards β

- The thermodynamic energy U , is the expectation value $\langle E \rangle$

$$U = \langle E \rangle = \sum_k N_k E_k = N \sum_k P(k) E_k$$

- which gives

$$\left(\frac{\partial U}{\partial \beta} \right)_{N,V} = N \sum_k E_k \frac{\partial P(k)}{\partial \beta}$$

- and for the entropy we had

$$S(U, V, N) = -Nk_B \sum_k P(k) \ln P(k)$$

$$\left(\frac{\partial S}{\partial \beta} \right)_{N,V} = -Nk_B \sum_k (\ln P(k) + 1) \frac{\partial P(k)}{\partial \beta}$$

- Because $\sum P(k) = 1$ $\sum \frac{\partial P(k)}{\partial \beta} = 0$

$$\left(\frac{\partial S}{\partial \beta} \right)_{N,V} = -Nk_B \sum_k \ln P(k) \frac{\partial P(k)}{\partial \beta}$$

towards β

- We use $P(k) = Q^{-1} e^{-\beta E_k}$ for the first $P(k)$

$$\begin{aligned} \left(\frac{\partial S}{\partial \beta} \right)_{N,V} &= -N k_B \sum_k \ln \frac{e^{-\beta E_k}}{Q} \frac{\partial P(k)}{\partial \beta} \\ &= -N k_B \sum_k (-\beta E_k - \ln Q) \frac{\partial P(k)}{\partial \beta} \\ &= N \beta k_B \sum_k E_k \frac{\partial P(k)}{\partial \beta} \end{aligned} \quad \left. \begin{array}{l} \text{red arrow} \\ \text{red arrow} \end{array} \right\} \sum \frac{\partial P(k)}{\partial \beta} = 0$$

- Coming back to the temperature, we get

$$\begin{aligned} T &= \left(\frac{\partial U}{\partial \beta} \right)_{N,V} / \left(\frac{\partial S}{\partial \beta} \right)_{N,V} \\ &= N \sum_k E_k \frac{\partial P(k)}{\partial \beta} / N \beta k_B \sum_k E_k \frac{\partial P(k)}{\partial \beta} \\ &= \frac{1}{\beta k_B} \end{aligned} \quad \boxed{\beta = \frac{1}{k_B T}}$$

Summary

- For a system of N particles and a constant energy E :
 - There are many different configurations possible
 - Every configuration has the same probability
 - Various configurations lead to the same distribution
 - The system will adopt the most likely distribution
 - The most likely distribution corresponds to the highest entropy, S
 - Maximize the number of possibilities to get a certain distribution, under the condition that N and E are constant
 - Leads to Boltzmann distribution, with the probability for any particle to be in state k : $P(k) = Q^{-1} e^{-\beta E_k}$
 - Q is the partition function
 - $\beta = 1/k_B T$

Summary

- Any microstate with constant N , V , E has equal probability
- Boltzmann linked the statistical entropy to
 - the number of different states
 - the number of microstates
 - the phase space integral
- From the microcanonical ensemble (NVE) we can derive all other ensembles:
 - Canonical ensemble (NVT)
 - Isothermal-isobaric ensemble (NPT)
 - Grand-Canonical ensemble (μVT)
 - ...

Backup slides

MATHEMATICAL BACKGROUND

Some maths – Stirlings approximation

- We will often need $N!$ for very large N
- Consider N as a continuous variable and $N!$ as a continuous function of N that can be differentiated with respect to N

$$\ln N! = \ln[N(N-1)(N-2)\dots\cdot 1]$$

$$= \ln \prod_{k=1}^N k = \sum_{k=1}^N \ln k \approx \int_1^N \ln x dx$$

use: $\ln(ab) = \ln a + \ln b$

replace sum by integral

$$= [x \ln x - x]_1^N$$

$$= N \ln N - N + 1$$

$$\approx N \ln N - N$$

for large N , we can neglect 1

Some maths – Lagrange multipliers

- We need to minimize or maximize a function $f(x_1, x_2, \dots, x_n)$: solve

$$df = \frac{\partial f}{\partial x_1} dx_1 + \frac{\partial f}{\partial x_2} dx_2 + \dots + \frac{\partial f}{\partial x_n} dx_n = 0 \quad \text{change } df \text{ in } f \text{ should be zero}$$

- If we want to find the solution, under specific conditions, we can formulate a constraint in the form $g(x_1, x_2, \dots, x_n) = 0$
- Then the easy solution to $df = 0$ by setting all $\partial f / \partial x = 0$ does not work, because $g(x_1, x_2, \dots, x_n)$ makes that the dx_i are dependent:

$$dg = \frac{\partial g}{\partial x_1} dx_1 + \frac{\partial g}{\partial x_2} dx_2 + \dots + \frac{\partial g}{\partial x_n} dx_n = 0 \quad \text{is zero because } g \text{ is constant}$$

- Instead, we solve this equation

$$\left(\frac{\partial f}{\partial x_1} + \lambda \frac{\partial g}{\partial x_1} \right) dx_1 + \left(\frac{\partial f}{\partial x_2} + \lambda \frac{\partial g}{\partial x_2} \right) dx_2 + \dots + \left(\frac{\partial f}{\partial x_n} + \lambda \frac{\partial g}{\partial x_n} \right) dx_n = 0$$

- Now, the dx_i are independent: set the terms in () to zero and solve for the Lagrange multiplier λ

Some maths – Lagrange multipliers

- Example
- Find the stationary points of the function $f(x,y)=4x^2 + 3x + 2y^2 + 6y$ under the constraint $g(x,y) = y - 4x - 2 = 0$

$$\frac{\partial f}{\partial x} = 8x + 3 \qquad \frac{\partial f}{\partial y} = 4y + 6$$

$$\frac{\partial g}{\partial x} = -4 \qquad \frac{\partial g}{\partial y} = 1$$

$$\left(\frac{\partial f}{\partial x} + \lambda \frac{\partial g}{\partial x} \right) dx + \left(\frac{\partial f}{\partial y} + \lambda \frac{\partial g}{\partial y} \right) dy = 0$$

$$(8x + 3 + \lambda(-4))dx + (4y + 6 + \lambda)dy = 0$$

- Gives a set of three equations with three unknowns:

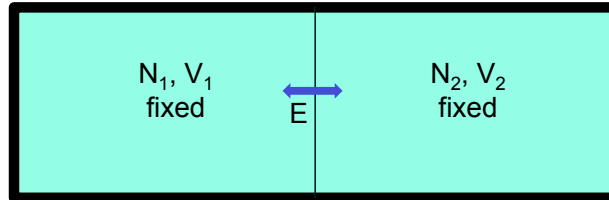
$$\left. \begin{array}{l} 8x - 3 - 4\lambda = 0 \\ 4y + 6 + \lambda = 0 \\ y - 4x - 2 = 0 \end{array} \right\} \Rightarrow \begin{array}{l} x = -59/72 \\ y = -23/18 \\ (\lambda = 8/9) \end{array}$$

Backup slides

LINK TO THERMODYNAMICS

Link to thermodynamics

- Let's consider two systems at constant N and V that can exchange heat.
- Together they form a closed and isolated system



- System 1: N_1, V_1 fixed; E_1 variable
- System 2: N_2, V_2 fixed; E_2 variable
- Two systems together $E = E_1 + E_2$ is constant
- what are the most probable value of E_1 and E_2 in the systems
- write these optimal values as E_1^* and $E_2^* = E - E_1^*$

Link to thermodynamics

- Solution is very similar to the individual particles earlier
- The probability that we have E_1 for system 1 (and also E_2 for system 2) is proportional to the number of possibilities

$$P(E_1) \propto \Omega_1(E_1)\Omega_2(E_2) = \Omega_1(E_1)\Omega_2(E - E_1)$$

- The maximum of this is

$$\frac{\partial}{\partial E_1} \Omega_1(E_1)\Omega_2(E - E_1) = \frac{\partial \Omega_1(E_1)}{\partial E_1} \Omega_2(E - E_1) + \Omega_1(E_1) \frac{\partial \Omega_2(E - E_1)}{\partial E_2} \cdot -1 = 0$$

$$\frac{\partial \Omega_1(E_1)}{\partial E_1} \Omega_2(E - E_1) = \Omega_1(E_1) \frac{\partial \Omega_2(E - E_1)}{\partial E_2}$$

$$\frac{1}{\Omega_1(E_1^*)} \frac{\partial \Omega_1(E_1)}{\partial E_1} \Big|_{E_1=E_1^*} = \frac{1}{\Omega_2(E - E_1^*)} \frac{\partial \Omega_2(E - E_1)}{\partial E_2} \Big|_{E_1=E_1^*}$$

$$\frac{\partial}{\partial E_1} \ln \Omega_1(E_1^*) = \frac{\partial}{\partial E_2} \ln \Omega_2(E - E_1^*)$$

Link to thermodynamics

- We already had that

$$S(U, V, N) = k_B \ln W(\mathcal{N}^*)$$

$$S_1(U, V, N) = k_B \ln \Omega_1(E_1^*)$$

- which means that the last line from the previous slide becomes

$$\frac{\partial}{\partial E_1} \ln \Omega_1(E_1^*) = \frac{\partial}{\partial E_2} \ln \Omega_2(E - E_1^*)$$

$$\frac{1}{k_B} \frac{\partial S_1(E_1^*)}{\partial E_1} = \frac{1}{k_B} \frac{\partial S_2(E - E_1^*)}{\partial E_2}$$

- And because $T = \left(\frac{\partial U}{\partial S} \right)_{N, V}$

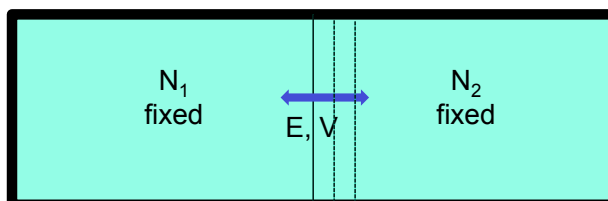
- we get $\frac{1}{k_B T_1} = \frac{1}{k_B T_2} \Rightarrow T_1 = T_2$

This will be an important finding when we introduce the canonical NVT ensemble

- So the two systems will be in thermal equilibrium

And mechanical equilibrium

- Let's consider two systems at constant N that can exchange heat and mechanical volume
- Together they form a closed and isolated system



- System 1: N_1 fixed; E_1, V_1 variable
- System 2: N_2 fixed; E_2, V_2 variable
- Two systems together $E = E_1 + E_2$ and $V = V_1 + V_2$ are constant

- what are the most probable values of E_1 and E_2 and of V_1 and V_2 in the systems?

- write these optimal values as E_1^* and $E_2^* = E - E_1^*$ V_1^* and $V_2^* = V - V_1^*$

Changing E and V

- Solution is very similar to the previous case
- The probability that we have E_1, V_1 for system 1 (and also E_2, V_2 for system 2) is proportional to the number of possibilities

$$P(E_1, V_1) \propto \Omega_1(E_1, V_1) \Omega_2(E - E_1, V - V_1)$$

- Setting

$$\frac{\partial}{\partial E_1} P(E_1, V_1) = 0$$

- leads to

$$\frac{1}{k_B} \frac{\partial S_1(E_1, V_1)}{\partial E_1} = \frac{1}{k_B} \frac{\partial S_2(E - E_1, V - V_1)}{\partial E_2}$$

$$T_1 = T_2$$

- Next step, also set

$$\frac{\partial}{\partial V_1} P(E_1, V_1) = 0$$

The volume derivatives

- So

$$\begin{aligned} \frac{\partial}{\partial V_1} \Omega_1(E_1, V_1) \Omega_2(E - E_1, V - V_1) = \\ \frac{\partial \Omega_1(E_1, V_1)}{\partial V_1} \Omega_2(E - E_1, V - V_1) + \Omega_1(E_1, V_1) \frac{\partial \Omega_2(E - E_1, V - V_1)}{\partial V_2} \cdot -1 = 0 \end{aligned}$$

- which we reshuffle to

$$\begin{aligned} \frac{\partial \Omega_1(E_1, V_1)}{\partial V_1} \Omega_2(E - E_1, V - V_1) &= \Omega_1(E_1, V_1) \frac{\partial \Omega_2(E - E_1, V - V_1)}{\partial V_2} \\ \frac{1}{\Omega_1(E_1, V_1^*)} \frac{\partial \Omega_1(E_1, V_1)}{\partial V_1} \Big|_{V_1=V_1^*} &= \frac{1}{\Omega_2(E - E_1, V - V_1^*)} \frac{\partial \Omega_2(E - E_1, V - V_1)}{\partial V_2} \Big|_{V_1=V_1^*} \\ \frac{\partial}{\partial V_1} \ln \Omega_1(E_1, V_1^*) &= \frac{\partial}{\partial V_2} \ln \Omega_2(E - E_1, V - V_1^*) \end{aligned}$$

Mechanical equilibrium

- Putting in the entropy definition

$$S_1(U, V, N) = k_B \ln \Omega_1(E_1^*)$$

- gives

$$\frac{\partial}{\partial V_1} \ln \Omega_1(E_1, V_1^*) = \frac{\partial}{\partial V_2} \ln \Omega_2(E - E_1, V - V_1^*)$$

$$\frac{1}{k_B} \frac{\partial S_1(E_1, V_1^*)}{\partial V_1} = \frac{1}{k_B} \frac{\partial S_2(E - E_1, V - V_1^*)}{\partial V_2}$$

- And because

$$T = \left(\frac{\partial U}{\partial S} \right)_N \quad P = \left(\frac{\partial U}{\partial V} \right)_N \quad \frac{P}{T} = \left(\frac{\partial S}{\partial V} \right)_N$$

- we get

$$\frac{P_1}{k_B T_1} = \frac{P_2}{k_B T_2} \Rightarrow P_1 = P_2$$

This will be an important finding when we introduce the isothermal isobaric NPT ensemble

- So the systems will be in thermal and mechanical equilibrium