

Optimizing Multiple Constraints

Note Title

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- Recall, in the method of Lagrange multipliers we studied, we were extremizing a function $f(x)$ subject to a constraint $g(x) = 0$
- What if we have more than one constraint?

Problem Let $f: \mathbb{R}^n \rightarrow \mathbb{R}$ be a C^1 function and $\phi_j: \mathbb{R}^n \rightarrow \mathbb{R}$; $j = 1, \dots, m$ be m C^1 functions. Let $S = \{x \in \mathbb{R}^n : \phi_j(x) = 0 \text{ for } j=1, \dots, m\}$. Which points in S extremize f over S ?

i.e. Extremize $f(x_1, \dots, x_n)$ subject to constraints
$$\begin{cases} \phi_1(x_1, \dots, x_n) = 0 \\ \vdots \\ \phi_m(x_1, \dots, x_n) = 0 \end{cases}$$

Key Idea: IF we only consider points x that satisfy the constraints, then $(x, f(x))$ is an extremum iff the constraints at that point do not allow movement in a direction that changes f

→ otherwise, we could move in that direction to make f bigger or smaller

→ in other words, $df = 0$ for all $dx = (dx_1, dx_2, \dots, dx_n)$

$$\begin{aligned} \text{where } df &\approx f(x + dx) - f(x) \\ &\approx \nabla f(x) \cdot dx \end{aligned}$$

- Consider the level sets of f at x
 - let $\{\hat{v}_L\} \subset \mathbb{R}^n :=$ the set of directions ($\|\hat{v}_L\| = 1$) in which we can move and still remain in the level set
 - i.e. these are the unit vectors for which $df = 0$

$$\therefore \forall \hat{v}_L \text{ in this set } \nabla f \cdot \hat{v}_L = 0 \quad (\text{I})$$

- Consider the constraints $\phi_j(x) = 0$, $j = 1, \dots, m$ at x
 - let $\{\hat{v}_c\} \subset \mathbb{R}^n :=$ the set of directions ($\|\hat{v}_c\| = 1$) in which we can move and still satisfy the constraint
 - since the constraints are $\phi_j = 0$ (a constant), we have just like for level sets $f = \text{constant}$ that the $\{\hat{v}_c\}$ are those directions for which $d\phi_j = 0 \quad \forall j$
 - or, $\nabla \phi_j \cdot \hat{v}_c = 0 \quad \forall j \quad (\text{II})$

"perpendicular to"
↓

(I) \Rightarrow Directions that don't change the value of f are $\perp \nabla f$
 $\Rightarrow \nabla f$ changes the value of f (provided $\nabla f \neq 0$)

(II) \Rightarrow All directions that still satisfy the constraints are $\perp \nabla \phi_j$, $j=1, \dots, m$

(I) + (II) A point \underline{x} is a constrained extremum of f iff the direction that changes f violates at least one of the constraints.

\Leftrightarrow The direction that changes f the most — namely, ∇f — violates at least one of the constraints

- a direction \hat{v} violates a constraint $\phi_i \Leftrightarrow \nabla \phi_i \cdot \hat{v} \neq 0$
- i.e. each constraint yields a "forbidden direction" given by $\nabla \phi_i$
- set of forbidden directions $\{\nabla \phi_i\}_{i=1}^m$

Then the direction that changes f violates at least one constraint

$$\Leftrightarrow \nabla f \in \text{span} \{ \nabla \phi_i \}_{i=1}^m$$

Lagrange multipliers
↓

$$\Leftrightarrow \nabla f = \lambda_1 \nabla \phi_1 + \dots + \lambda_m \nabla \phi_m \quad \text{for } \lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R}$$

$$\text{or} \quad \nabla f - \sum_{i=1}^m \lambda_i \nabla \phi_i = 0$$

Thm (Method of Lagrange Multipliers) Extremal points \underline{x} of f on the constraints $\phi_1, \phi_2, \dots, \phi_m$ satisfy

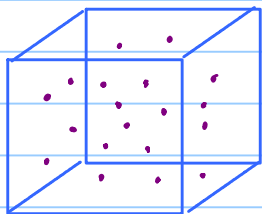
$$\nabla f(\underline{x}) - \sum_{i=1}^m \lambda_i \nabla \phi_i(\underline{x}) = 0 \quad \text{and} \quad \phi_j(\underline{x}) = 0 \quad \text{for } j=1, \dots, m$$

for some $\lambda_1, \lambda_2, \dots, \lambda_m \in \mathbb{R}$ called Lagrange Multipliers.

- Notes:
- we still also need to check points where $\nabla \phi_j(\underline{x}) = 0$
 - we still also need to check endpoints of the constraints.

Example: Statistical Mechanics

very large, $\sim 10^{26}$
↓



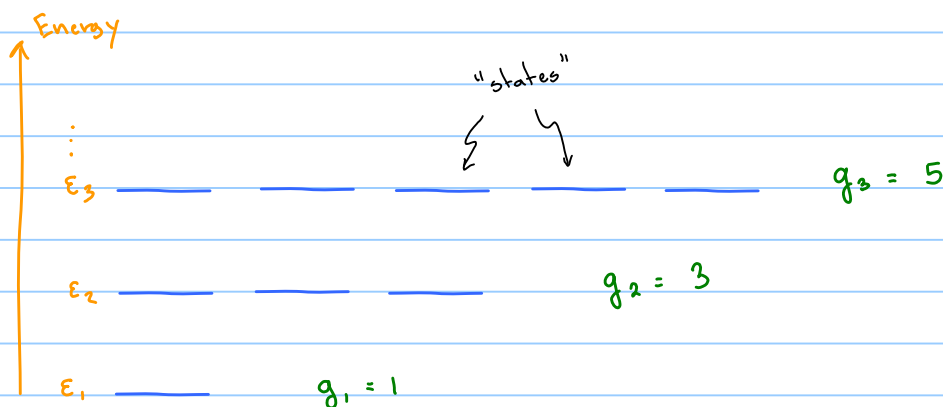
- Suppose we have some collection of N particles where each particle can have as its energy one of

$$0 < \epsilon_1 < \epsilon_2 < \dots < \epsilon_n$$

and that there are g_j states available to a particle if it has energy ϵ_j

(i.e. there are g_j ways for a particle to have energy ϵ_j)

Pictorially,



- The configuration of the system is specified by the distribution of the particles among the system's energy levels
i.e. the numbers $N_j := \#$ of particles with energy ϵ_j

- Notice, we must have $\sum_{j=1}^n N_j = N$ — (1)

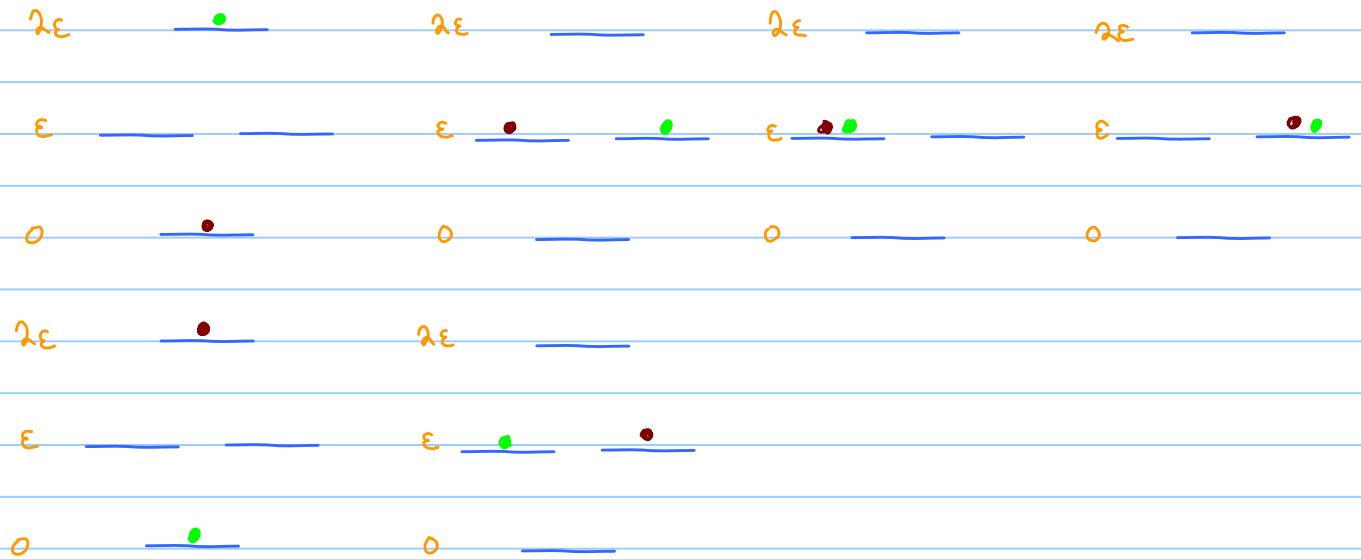
- We pose the following question:

Suppose our system has a total energy $U = \sum_{j=1}^n N_j \epsilon_j$. — (2)

What are the values taken by the N_j , $j=1, \dots, n$?
Or, what is the configuration the system assumes?

• Notice, there be many different ways to produce a ^{given} configuration, or macrostate (N_0, \dots, N_n)

ex $N=2$, $U=2\varepsilon$, possible energies $\varepsilon_0=0$, $\varepsilon_1=\varepsilon$, $\varepsilon_2=2\varepsilon$
 $g_0=1$, $g_1=2$, $g_2=1$. Then, possible states are



→ 6 microstates

(Note: here we are assuming the particles are distinguishable)

Defⁿ $\omega(N_0, \dots, N_n) := \#$ of microstates that produce the macrostate (N_0, N_1, \dots, N_n)

So, in the above, eg. $\omega(N_0=0, N_1=2, N_2=0) = 4$
 $\omega(N_0=1, N_1=0, N_2=1) = 2$
 $\omega(\text{anything else}) = 0$

- Statistical mechanics postulates that how the particles in a system arrange themselves is completely random and that each microstate is equally likely

⇒ The most likely, or equilibrium configuration (N_1, N_2, \dots, N_n) is that for which $\omega(N_1, N_2, \dots, N_n)$ is maximal.

So, to answer the question,

Maximize $\omega(N_1, \dots, N_n)$ subject to the constraints

$$(1) \quad \sum N_i = N \quad [\text{conservation of particle number}]$$

$$(2) \quad \sum N_i \varepsilon_i = U \quad [\text{conservation of energy}]$$

Solution For N distinguishable particles with energy levels $\varepsilon_1, \dots, \varepsilon_n$ and degeneracies g_1, \dots, g_n , one can show using combinatorics that

$$\omega(N_1, \dots, N_n) = N! \prod_{i=1}^n \frac{g_i^{N_i}}{N_i!} \quad (3)$$

- since \ln is a monotonically-increasing function, let's maximize $\ln \omega$ instead of ω

$$\ln \omega = \ln N! + \sum_{i=1}^n [N_i \ln g_i - N_i!]$$

$$\approx N \ln N - N + \sum [N_i \ln g_i - N_i \ln N_i + N_i]$$

$$= N \ln N + \sum [N_i \ln g_i - N_i \ln N_i]$$

Stirling's approximation

$$\ln x! \approx x \ln x - x$$

for large x

(recall $N \sim 10^{26}$)

Constraints (1): $\phi_1(N_1, \dots, N_n) = \sum N_i - N = 0$

(2): $\phi_2(N_1, \dots, N_n) = \sum N_i \varepsilon_i - U = 0$

Now, we look for (N_1, \dots, N_n) satisfying

$$\nabla(\ln \omega) + \lambda_1 \nabla \phi_1 + \lambda_2 \nabla \phi_2 = 0$$

$$\Rightarrow \frac{\partial \ln \omega}{\partial N_j} + \lambda_1 \frac{\partial \phi_1}{\partial N_j} + \lambda_2 \frac{\partial \phi_2}{\partial N_j} = 0 \quad \text{for } j=1, \dots, n$$

We have $\frac{\partial \ln \omega}{\partial N_j} = \ln g_j - \ln N_j - 1$

$$\approx \ln g_j - \ln N_j \quad \text{for large } N_j, g_j$$

$$\frac{\partial \phi_1}{\partial N_j} = 1$$

$$\frac{\partial \phi_2}{\partial N_j} = \epsilon_j$$

$$\Rightarrow \ln g_j - \ln N_j + \lambda_1 + \lambda_2 \epsilon_j = 0$$

$$\ln \left(\frac{N_j}{g_j} \right) = \lambda_1 + \lambda_2 \epsilon_j$$

$$\boxed{\frac{N_j}{g_j} = e^{\lambda_1 + \lambda_2 \epsilon_j}} \quad (*)$$

This is the famous Maxwell-Boltzmann distribution - it tells us the values of N_j , i.e. the system's configuration.

- To eliminate λ_1 and λ_2 , we would need to use the constraints.

From $\sum N_j = N$ and $(*)$,

$$\sum N_j = \sum g_j e^{\lambda_1 + \lambda_2 \epsilon_j} = e^{\lambda_1} \sum g_j e^{\lambda_2 \epsilon_j} = N$$

The "partition function"

$$\Rightarrow e^{\lambda_1} = \frac{N}{\sum g_j e^{\lambda_2 \epsilon_j}} = \frac{N}{Z} \quad \text{where } Z = \sum g_j e^{\lambda_2 \epsilon_j}$$

- Solving for λ_2 is typically done using certain results from thermodynamics - one could use the second constraint, but it's pretty messy. A good exercise, however!

Propⁿ $\lambda_a = -\frac{1}{kT}$ where T : system temperature
 k : Boltzmann's constant

Then, the equilibrium configuration of the system is given by

$$\frac{N_j}{g_j} = \frac{N}{Z} e^{-\epsilon_j/kT}$$

The Maxwell-Boltzmann Distribution.