CELL MECHANICS

LECTURE 3 2. Physical principles

2.1. Forces at molecular and cell level

- Physical forces and their magnitudes at the single-molecule level
- Modeling complex mechanical devices as protein machines by using three elements: Spring, Dashpot, Mass; example: Mass, Stiffness and Damping of Proteins

2.2. Thermal forces, diffusion, and chemical forces

- Boltzmann Distribution Law and the Principle of Equipartition of Energy
- Diffusion equation Einstein relation Stokes law
- Autocorrelation function and Power Spectrum
- The effect of force on the equilibria and rate of chemical reactions
- Example of single molecule force spectroscopy experiments unbinding, unfolding

Reference: Book_Howard_Ch_4

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Lecture 2. Forces at molecular and cell level

By considering three mechancial elements – mass, damping, spring - we introduced some of the mechanical concepts required to understand how forces influence protein and cells.

movement of a cell / bacterium or a protein through a liquid, inertia of a protein (100 kDa), measured as time to stop ($\tau = 3 ps$)

protein that undergoes a global conformational change ($\tau = 15 \text{ ns}$)

Vibration of chemical bonds ex. H-Cl, f= 89.6 10¹² Hz; mass m~1.63 10⁻²⁷ kg \rightarrow k = m ω^2 = 517 N/m – very stiff

The mass and spring with damping illustrate that system can respond to mechanical forces int two ways:

they can **oscillate** or they can move **monotonically.**

• Due to the small size of proteins, the viscous forces are preponderent over the inertial forces \rightarrow global motions of proteins, especially less rigid ones, are overdamped \rightarrow they creep rather than oscilate when subject to forces

$$
\frac{\gamma^2}{4mk} \cong 25 \frac{\eta^2}{\rho E L^2} > 1 \qquad k = \frac{AE}{L}
$$

- The motion of long, thin cytoskeletal filaments are also overdamped, due to their large aspect ratios, this in turn causes the motion of the cells to be overdamped.
- The rigidity of cytoskeleton proteins as actin, tubulin, keratin is similar to that of hard plastics but less than that of other materials such as glass or metal, because proteins are held together by weak Van der Waals bonds
- The rigidity of protein machines/motors undergoing large conformational changes as they transduce chemical energy into mechanical work is expected to be much less than that of structural proteins
- As proteins move and change shape, they experience damping forces from the surrounding fluid as well from the internal friction. These forces arise from the rapid making and breaking of bonds.

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Thermal forces and diffusion 6

In addition to mechanical forces, proteins and cells are subject to **thermal forces**, arising from collisions with water and other molecules in the surrounding fluid

Thermal forces → **thermal energy** → **thermal / Brownian motion**

The magnitude of thermal energy is in the range of the energies of chemical reactions driving biological processes, which are just a little bit higher than thermal energy \rightarrow thermal fluctuations are necessary for proteins to reach their transition states

Molecular machines operate in diffusive environment, differently from macroscopic machines of our everyday world

Boltzmann Distribution Law

describes how the probability of a molecule having a certain energy depends on the surrounding temperature

Principle of Equipartition of Energy

states how much thermal energy a molecule has at a certain temperature

if a particle (or a group of particles) is in thermal equilibrium, the *probability p_i* of finding the particle in the state i, characterized by <u>*energy U_i* is given by:</u>

$$
p_i = \frac{1}{Z} \exp\left[-\frac{U_i}{kT}\right]
$$

where:

$$
Z = ct = \sum_{i} exp\left[-\frac{U_i}{kT}\right]
$$
 is the partition function ($\sum p_i = 1$)

- *K* is the Boltzmann constant, $K = 1.381 \times 10^{-23}$ [J K⁻¹]
- is the absolute temperature.

- Boltzmann distribution allows to calculate the probability of observing a system at finite temperature in any particular microstate.
- The probability only depends on the energy (free energy) of the state.

Molecules in a two-state energy landscape:

Anticipating: Protein unfolding – free energy landscape

A one-dimensional free energy diagram allowing for single unfolding pathway – transition path.

The extension x represents the unfolding reaction coordinates

A two-dimensional free energy diagram allowing for multiple unfolding pathways.

x1 and x2 represent generalized unfolding reaction coordinates.

Boltzmann's Law 9

For T= 298.15 K (Tc= 25 C) the energy KT = 4.116×10^{-21} [J] \rightarrow 1 KT \approx 4.1 [pN nm] Thermal energy KT is a convenient energy unit for processes at molecular and cellular level Comparison with other biologically relevant energies:

Planck constant *h= 6.6 x 10-34 [J s]*

Boltzmann's law is very general

The energy could correspond to the particle's potential energy (gravitational, elastic, or electrical), kinetic energy,

or other energy associated with its phase, or electronic or chemical state.

Note:

the state of a particle (or group of particles) is specified by the position and velocity of the constituent atoms as well their electronic states.

Boltzmann's law is fundamental because:

One can use it to define equilibrium and temperature:

- A system is at **equilibrium** if Boltzmann's law holds
- The **temperature** is defined as the corresponding constant in the exponent of the Boltzmann's law formula

Boltzmann's law is a very important physical law in biology and chemistry.

1. Earth's athmosphere

Knowing that the density of molecules in a gravitational field falls exponentially with the height, estimate the Earth's atmosphere scale height, SH.

SH **:** the height for which the density falls by 1/e (= 37%) Consider the gravitational potential energy: $U = mgh$, of a particle of mass m, at a height h above the Earth's surface. Molar mass for oxygen MM=32 g/mol, 1 mole has \approx 6.022 \cdot 10²³ units (Avogadro's number) **A:** SH ~ 7.5 km

2. Settling of beads

Same problem, considering glass microspheres of diameter d= 200 nm, (mass density of glass: 2 g/cm³) **A:** SH ~ 100 µm

3. Analytic centrifugation

Measuring the mass m of a protein with the analytic centrifuge.

U= (m-m^w)ach - potential energy;

m-m^w – additional mass over that of the displaced solvent (water);

h – height above the bottom of the centrifuge tube; a_c – centrifuge acceleration.

One measures the height SH10 for which the density of protein falls by $1/e^{10}$ (exp term vanishes)

 \rightarrow *U*₀ \approx 10 kT \rightarrow *m* = m_w + 10 kT/(a_c SH)

Exp values: ac $\sim 10^3$ g ; SH10 \sim 10 mm \rightarrow **m**= ?

$$
p_i = \frac{1}{Z} \exp\left[-\frac{U_i}{kT}\right]
$$

Homework

4. Nernst equation

Considering a set of molecules with charge *q* that are free to equilibrate between two compartments at electrical potential 0 and V volts, Find the ratio of the concentration of molecules in the two compartments.

A:

$$
\frac{C_V}{C_0} = \frac{p_V}{p_0} = exp\left[-\frac{U}{kT}\right] = exp\left[-\frac{qV}{kT}\right]
$$

 $q= 1.6 \times 10^{-19}$ C

At room temperature K= 300 K, kT/q = 25.6 mV

- -> for each 25.6 mV increase in voltage, the concentration of monovalent cations decreases *e*-fold.

The average thermal energy of a molecule (or system of molecules)

Example

Suppose a molecule is at equilibrium in an **energy landscape** *U(x)*, that varies with position *x* (but not with time);

e.g. the molecule could be connected to a spring with potential energy:

U(x)= ½ k x²

Due to thermal agitation, the molecule is constantly changing position.

Aim:

Calculate the statistical properties of the molecule's position:

mean, mean squared, variance

The statistical properties of a molecule's position, such as its mean or its variance can be calculated in two ways:

1. Follow the molecule over a long period of time T, and measure its time-averaged **mean** position or **mean-squared** position :

mean
$$
\langle x \rangle_T = \frac{1}{T} \int_0^T x(t) \cdot dx
$$
 mean-squared $\langle x^2 \rangle_T = \frac{1}{T} \int_0^T x^2(t) \cdot dx$ variance $\sigma_x^2 = \langle (x - \langle x \rangle)^2 \rangle = \langle x^2 \rangle - \langle x \rangle^2$

2. Use the Boltzmann's law to calculate **the probability p(x)** of finding the molecule at **position x** and then calculate the expected values E(x) of the position or position squared according to:

$$
E(x) = \int_{-\infty}^{\infty} x p(x) \cdot dx
$$

$$
E(x^2) = \int_{-\infty}^{\infty} x^2 p(x) \cdot dx
$$

If we measure for a long enough time, then the estimates of the average position should agree:

$$
\langle x \rangle \equiv \langle x \rangle_{\infty} = E(x) \qquad \qquad \langle x^2 \rangle \equiv \langle x^2 \rangle_{\infty} = E(x^2)
$$

In this way we can **relate measurements** (time averages) **to the expectations**, based on Boltzmann's law.

The Equation above is the link between experiments and theory!

It holds generally for any function of x : E[f(x)]=<f(x)>. In particular it holds for the variance of σ_x^2 of x.

This approach can be used to calculate the average energy of a molecule.

For instance, for the molecule attached to a spring, the average energy is:

 $\langle U \rangle = \frac{1}{2} \kappa \langle x^2 \rangle = \frac{1}{2} \kappa \int x^2 p(x) \cdot dx = \frac{1}{2} kT$

using Boltzmann's law for p(x):

$$
p(x) = \frac{1}{Z} \exp\left[-\frac{U(x)}{kT}\right]
$$

The result above is remarkable because the average energy <U> does not depend on the stiffness of the spring ! It only depends on the temperature *T* **!**

This is a special case of a general theorem known as the **Principle of Equipartition of Energy** which states that if the energy of a molecule depends on the square of a parameter (such as position or speed), the mean energy associated with the degree of fredom measured by the parameter is:

$$
\langle \boldsymbol{U}\rangle = \frac{1}{2} \boldsymbol{K} \boldsymbol{T}
$$

Another example of the principle is **that the average kinetic energy** of a molecule (in one direction) with mass m is:

$$
K.E. = \frac{1}{2}m\langle v^2 \rangle = \frac{1}{2}KT
$$

If there are more degrees of freedom, that are independent, then each degree of freedom contains ½ KT of energy.

E.g. : the velocities of a molecule in x, y, z directions are independent for 3 degrees of freedom.

Thus, the **total kinetic energy** is: **3/2 kT**.

The **root mean-square speed, v_{rms}**, of a molecule in three dimensions is therefore:

$$
v_{rms} = \sqrt{\langle v^2 \rangle} = \sqrt{\frac{3KT}{m}}
$$

Examples of v_{rms} at 25 C, for:

- Water molecule, v_{rms} = 640 m/s
- Protein, 100 kDa, v_{rms} = 8.6 m/s
- Bacterium of volume $1 \mu m^3$ = 3.5 mm/s

The Principle of Equipartition of Energy is generally true only if the energy dependence is quadratric.

If, for instance, $U(x) \sim x \implies \langle U \rangle = KT$ (and not 1/2 KT)

It breaks down also if *KT* is small compared to energy levels between different quantum states.

For proteins at room temperature thermal energy is large compared to the vibrational energy levels because proteins are relatively soft materials

(appendix 4.1 for details)

Thermal energy KT ~ 4 x 10⁻²¹ J while vibrational energy h v ~ 10⁻²² J (ω ~10¹² Hz, h ~ 6.6 10⁻³⁴ m²kg/s)

 \rightarrow hy < KT

→ **the principle of equipartition of energy applies to elastic deformation of proteins**

Molecular collisions cause **Brownian motion** and **diffusion**.

These are forms of random motion that are characterized by frequent, abrupt changes in direction.

In **Brownian motion**, a particle **does not have a specific direction** to travel → it will move in all directions.

In **diffusion** the particles will travel from a high concentration to a low concentration. Therefore, **they have a direction**.

However, the particle movement is random in both scenarios.

Diffusion plays a crucial role in many physical and chemical processes at microscopic scales.

- Einstein (deserved, not awarded) Nobel prize award for 'elucidating the molecular mechanism of Brownian motion'
- Perin and Svedberg Nobel prize measurements of the diffusion of micron-sized particles, confirmed Einstein's theory and allowed the measurement of Boltzmann's constant K and the determination of the Avogadro number N_{A} .
- Brownian motion confirmed the atomic theory of gases and liquids and bridged the gap between visible objects and invisible molecules.

In the presence of **a concentration gradient**, the molecules moving in random directions tend to move, in average, **from areas of high concentration to areas of low concentration**.

-> gradual change ->

few

The prediction, confirmed experimentally, is that the **concentration flux** *J(x)*,

lots

which is the rate of movement of molecules per unit area,

is proportional to the concentration gradient *dc/dx*:

Fick's first law of diffusion

$$
J(x) = -D \frac{dc(x)}{dx}
$$

D – diffusion coefficient

Diffusion equation

The **diffusion equation** (Fick's 2nd law of diffusion) can be derived from the 1st law:

The **change in concentration** over time at any point **equals** the negative of **the flux gradient** at that point**:**

$$
\frac{\partial c(x,t)}{\partial t} = -\frac{\partial J(x,t)}{\partial x}
$$

Note:

if the system is in steady – state (*dc/dt=0*), then the concentration flux is the same everywhere in the solution (*dJ/dx=0*). Conversely, if the flux does not change from one position to another, then the concentration does not change with time.

Substituting equation above into Fick's 1st law one gets:

$$
J(x) = -D \frac{dc(x)}{dx}
$$

Diffusion equation

$$
\frac{\partial c(x,t)}{\partial t} = D \frac{\partial^2 c(x,t)}{\partial x^2}
$$

Rather than the concentration *c(x,t)* of a large number of molecules, we can think about single molecules and want to know the probability *p(x,t)* of finding a molecule at position *x* at time *t.*

Because **the probability is proportional to the concentration** (it is the concentration divided by the total number of molecules) and because **differentiation is a linear operator**, it follows that also the **probability** *p (x,t)* **satisfies the diffusion equation**:

$$
\frac{\partial p(x,t)}{\partial t} = D \frac{\partial^2 p(x,t)}{\partial x^2}
$$

Boltzmann's law allows to derive an expression that relates the **diffusion coefficient** to the **drag coefficient (Einstein relation)**

Suppose that an external force, *F(x)*, acts on a diffusing molecule.

The force will cause the molecule to move with a *velocity v(x)= F(x) / ɣ .*This 'drift' velocity is an average speed superimposed on the diffusive motion. The external force increases the flux by *v(x) c (x,t)* or by *v(x) p(x,t),* if we are thinking of the **probability flux***, j(x)*:

$$
J(x) = -D \frac{dc(x)}{dx} \qquad j(x) = -D \frac{dp(x)}{dx} + \frac{F(x)}{\gamma} p(x)
$$

Einstein relation

Thus, in the presence of a force, the probability satisfies the equation (derived from diffusion equation):

$$
\frac{\partial p}{\partial t}(x,t) = D \frac{\partial^2 p}{\partial x^2}(x,t) - \frac{\partial}{\partial x} \left[\frac{F(x)}{\gamma} p(x,t) \right]
$$

This equation is known as the **forward diffusion equation** or the **Fokker-Planck equation** and describes diffusion with drift.

If the system is in **equilibrium**, the probability does not change with time and the **F-P equation** can be solved in *p(x)*. (See Appendix 4.2)

Comparing the solution of the Fokker-Planck with the Boltzmann's law, it is found that the flux must be equal to zero everywhere, and that the diffusion coefficient is related to the drag coefficient by:

$$
D = \frac{KT}{\gamma}
$$
 Einstein relation

relates the **diffusion coefficient** of a molecule to its **drag coefficient**

Einstein relation allows to estimate the **diffusion coefficient** from the **size** of the particle and the **viscosity** of the solution.

Conversely, knowledge of the viscosity and the diffusion coefficient permits an estimate of the size of the particle.

Example: Diffusion of ions. Consider sodium ion Na+ in water.

The diffusion coefficient for an ion at room temperature (25 C) is D= 1.33×10^{-9} m²/s.

From the Einstein relation it results an apparent radius $r= 1.8$ Å, which is about two times the ionic radius of 0.95 Å measured in crystals.

Rule of thumb:

a diffusion coefficient D= 10^{-9} m²/s = 1μ m²/ms, corresponds to the diffusion of a small ion for about 1 μ m in 1 ms.

Diffusion vs Osmosis 24

Diffusion is the movement of **particles** from an area of higher concentration to lower concentration. The overall effect is to equalize concentration throughout the medium.

Osmosis is the movement of solvent particles across a semipermeable membrane from a dilute solution into a concentrated solution. The **solvent** moves to dilute the concentrated solution and equalize the concentration on both sides of the membrane.

Facilitated diffusion is the movement of hydrophilic molecules or ions across the plasma membrane. The difference is, the diffusion is facilitated by transport proteins (i.e. carrier proteins or pore proteins). The utility of the diffusion equation is that it allows one to calculate **how quickly, on average, it takes for a molecule to diffuse through a certain distance.**

This information can be used to evaluate **the efficiency of diffusion as a transport process within cells.**

Furthermore, with the aid of the Fokker-Planck equation, we can calculate the time that it takes for a molecule to diffuse against an applied force. One can then gain insight into how forces affect chemical rates.

Solutions of the diffusion equation for some particular cases that are relevant to cellular and molecular mechanics.

Solutions of the diffusion equation for some particular cases that are relevant to cellular and molecular mechanics.

The diffusion from a point source

If a molecule is released at the origin and allowed to diffuse in one dimension, then the probability of finding it at position x at time t is:

$$
p(x,t) = \frac{1}{\sqrt{4\pi Dt}} \exp\left[-\frac{x^2}{4Dt}\right] \qquad t>0
$$

How far, on average, does a molecule diffuse in a given time ?

root-mean-square displacement : $x_{rms} = \sqrt{2Dt}$ (Standard deviation)

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The diffusion from a point source

Another, relevant question is**:** how long on average, does it take a molecule to diffuse through a given distance ?

first-passage time:
$$
t = \frac{x_0^2}{2D}
$$

First passage time is relevant because it allows to calculate the rate of a process that is limited by diffusion.

The **diffusion-limited rate**, *kdl* is the reciprocal of the **first-passage time, t:** *kdl = 1/t*

In the **absence of an external force**, the first-passage time for one-dimensional diffusion through a distance x₀ is:

The first-passage time can be calculated by solving the diffusion equation for the particular geometry of the problem.

Diffusion as a cellular transport mechanism

Case study: Evaluate if diffusion might be a feasible mechanism to transport molecules and organelles in the cell.

Let us consider a globular protein, a potassium ion K+, and an organelle (mitochondrion).

How long it takes for these three particles to propagate different distances in the cell ? Size of cell: max 100 µm.

Protein and ion diffusion are efficient, the organelle diffusion is very slow.

Actually, the organelle diffusion is even slower, because the cytoplasm is like a gel with mesh size of about 50 nm. Organelles larger than 50 nm are almost immobile.

Motor proteins are required to move organelles from one place to another.

On the other side, the low mobility of the organelles is benefic: large organelles stay where they are, and the internal structure of the cell will be reasonable stable.

KT

 $6\pi\eta r$

 $t=$

 x_0^2

 $D=$

 $2D$

However, there are several **"microscopic" details of diffusive motion** that are important to answer questions as:

- 1. How long, on average, will a free molecule keep moving in one direction before thermal forces randomize its direction of motion ?
	- i.e. which is the **persistence length** (or **correlation time**) **of the velocity** ?

- 2. How long, on average, will it take for a molecule in a potential well to explore the different energy levels ? In particular, how long will the molecule spend at each energy level ?
	- i.e. what is **the persistence time of the position** ?

3. What are the **amplitude** and **statistical properties** of the thermal forces ?

Diffusion of a free protein vs diffusion of a tethered protein

The root-mean-square velocity: $v_{rms} = \sqrt{\frac{3KT}{m}}$ $\frac{KT}{m} \approx 8.6 \frac{m}{s}$ $\frac{n}{s}$; The time constant $\tau_i = \frac{m}{\nu}$ $\frac{m}{\gamma} \approx 3 \ p s$ (with $\gamma \cong 60 \ pN \cdot s/m$) is the **correlation time of the velocity**

The corresponding **persistence length**: $l = v \cdot \tau = 0.24 A!$

Even if the speed of molecule is large, the high damping it experiences in water opposes to inertia and after just a fraction of an A it changes direction.

Protein attached to a spring of stiffness *k= 1pN/nm* The root-mean-square displacement (using Th Equipartition Energy)

$$
x_{rms} = \sqrt{\langle x^2 \rangle} = \sqrt{\frac{\kappa r}{k}} \approx 2 \; nm \; ;
$$

The time constant $\tau_p = \frac{\gamma}{k}$ $\frac{y}{k} \approx 60 \text{ ns} = \tau_i \cdot 2 \cdot 10^4$

This is the **correlation time of the protein's position**, i.e. the time it takes to the protein to relax to a new position.

For times t< τ_p , the protein is quite near the same position; when t>>τ_ρ, the protein's position is uncorrelated and the probability of finding the protein in a certain position depends only on its potential energy and not on time.

RECONSIDERING the Motion of Combinations of Mechanical Elements in presence of thermal forces

A) DASHPOT and MASS. Model for the movement of a protein through a liquid

 $\overline{\gamma}$

(A)
\n
$$
F_d = -\gamma v
$$

\n γ
\n γ

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J. Howard – Book, Ch. 2

What if
$$
F = \phi
$$
?
\n
$$
m\frac{dv}{dt} + rv = \phi
$$
?
\n
$$
m\frac{dv}{dt} + rv = \phi
$$

\n
$$
\phi = \frac{1}{2} ar
$$

\n
$$
\frac{dv}{dt} + rv = \phi
$$

\n
$$
\frac{dv}{dt} + \frac{dv}{dt} = \phi
$$

\n
$$
\frac{dv}{dt} + \frac{dv}{dt} = \phi
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\n
$$
\frac{dv}{dt} + \frac{dv}{dt} = \phi
$$

\n
$$
\frac{dv}{dt} = \frac{1}{2} ar
$$

\n
$$
\frac{dv}{dt}
$$

 E_5 . of motion; $\left[\frac{dv}{dt} - \frac{v}{u}v + \frac{1}{m}\zeta(t) \right]$ Langevin Canditions $\langle \xi(t)\rangle = 0$ $\langle \xi(t_1) \xi(t_2)\rangle = g \cdot \delta(t_1-t_2)$
for $\xi(t)$ first monuent $\zeta(t)$ -delta function -todicates there is no correlation between impacts in any distinct
time intervals dt, and dt, correlation time is infinitely short g - is the amplitude of the fluctuation force, $22kT$ Pluctuation dissipation theorem

 22272 Pluctuation dissipation theorem $\overline{f\Delta T}$ • The aw plitude of the therwal force depends only on
the drag coefficient (foiction dissipation).
• FBT expresses the balance between friction which tends
to drive any system to a completely dead state and noise Which tends to keep the system alive. 6 This balance is required to have a therwal equilibrium state at long times. Ref: Ch. 6 - Brownian Mation: Langenn Equation

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http://physics.gu.se/~frtbm/joomla/media/mydocs/LennartSjogren/kap6.pdf

https://web.stanford.edu/~peastman/statmech/friction.html

MASS and SPRING with DAMPING

Mechanical model of a protein undegoing a large scale conformational change that is damped by the surrounding fluid, and possibly by internal viscosity.

 $F = \phi$, $ud^2x + g^2 du' + kx = \xi(t)$ $\frac{66}{7}$ -stochadic => n can be described only by
its statistical properties;

Autocorrelation function, WHY ?

Using the autocorrelation function Rx (E): $R_{\chi}(s) = \chi(t) * \chi(t) = \lim_{T \to \infty} \left\{ \frac{1}{T} \int_{-\infty}^{T} \chi(t) \cdot \chi(t-s) \cdot dt \right\} =$ = $\langle x(t) - x(t-1) \rangle$ The autocorrelation at delay & is calculated by multiplying and averaging over all times. T^2 $\left(\frac{R_x(\tau)}{\tau}\right)$. $R_{x}(0)$ $\tau(x)$ *variance* -0.5

Autocorrelation function, WHY ?

Autocorrelation function the drispes the eg. of motion; me d'Rx + se dRx + kRx = procourse It means that AI has the same form as the response Consequence: to estimate the molecule's moleculor properties fit it to a prodicted model, i.e. m, je, k are pitting por aux eters.

38 *Fourier transform and power spectrum*

Use the Fourier analysis of the signal Power Spectorum: PS $\begin{array}{rcl} P5 &= \frac{7}{2} \{ A \overline{F} \} & \hat{F} - \text{Fourier} & \text{transfo} & \text{if} \\ S\sigma : & -\text{we measure the thermal motion of the molecule} \\ & -\text{colculate the PS} & PS = \frac{1}{2} \{ A \overline{F} \} = \overline{f}(x) \{ \overline{f}(x) \} \end{array}$ $PS = f - \frac{1}{2}AF = f(x) \cdot f(x)$ $=$ $\frac{2}{300}$ - compose PS with PS of a predicted model Similar approach as in X-ray cristallography.

Schematic of a microbead in an optical trap – otpical tweezers 39

Frequency (Hz)

ξ(t)

MASS and SPRING with DAMPING model

Model also for a protein undegoing a large scale conformational change that is damped by the surrounding fluid, and possibly by internal viscosity.

Determine the stiffness of the optical trap k:

- measure the fluctuations of the bead x(t) over a period of time (e.g. 2 s) at high frequency (10 kHz)
- calculate the power spectrum PS(f) and fit it with a Lorenzian function of parameters *fc*, *γ*

Lorenzian function

$$
S_x(f) = \frac{k_B T}{\pi^2 \gamma (f_C^2 + f^2)}
$$

$$
f_{\rm C} = k/2\pi\gamma
$$

See the lesson on Optical Tweezers for a more detailed discussion

Summary

$$
p_i = \frac{1}{Z} \exp\left[-\frac{U_i}{kT}\right]
$$

Boltzmann's law gives the probability that a molecule is in a given energy state.

Boltzmann's law has some corollaries:

1. Principle of the Equipartition of Energy which says that each degree of freedom of the molecule has ½ KT of energy associated with it. KT= 4.1 pN nm (room temperature).

An example of this principle is that the average of the potential energy of a (molecular) spring is ½ KT.

Another example is that the average kinetic energy of a molecule is 3/2 KT.

The randomly directed collisions with the surrounding molecules cause particles to move randomly (Brownian motion) and diffuse . The statistical properties of diffusing particles can be derived by solving the diffusion equation.

2.The diffusion coefficient D is related to the drag coefficient γ via the equation D= KT/ γ – Einstein equation. A detailed analysis of diffusion shows that the velocity persists (or is correlated) over times shorter than m/ γ . For a protein this time is of the order of ps after which time the protein is likely to be moving in a different direction and with a different speed. If the protein is tethered by a spring of stiffness k, the position is correlated over times shorter than γ /k, which ranges from 1 ns to 1 us for spring constants k between 0.016 and 16 pN/nm.

Langevin equation.

Using the Fourier analysis it is possible to deduce the molecular properties of a mechanical system, such as stiffness, damping and mass, from the analysis of the thermal motion.

From a biological point of view, it is interesting to consider **diffusion in the presence of an external force**.

e.g. how long does it take for a molecule to diffuse over an energy barrier at *x= x⁰* ?

Considering the force constant, the potential energy is *U(x)= - Fx*, and the **first-passage time**:

Time for a 100 kDa protein to diffuse 8 nm in presence of a constant force.

$$
t = 2\left(\frac{x_0^2}{2D}\right)\left(\frac{kT}{Fx_0}\right)^2 \left\{\exp\left(-\frac{Fx_0}{kT}\right) - 1 + \frac{Fx_0}{kT}\right\}
$$

Uphill: against an opposing force the first-passage time t is long (and the corresponding diffusion-limited rate is reduced)

Downhill: in direction of the force, t is short.

The external force : $F = -kx$.

How long does it take for a molecule to diffuse over an energy barrier at *x= x⁰* ?

When the force opposes the motion: *F= -k x ,* the potential energy is *U(x)=- k x²* , and the **first-passage time**:

$$
t_{\rm K} = \tau \sqrt{\frac{\pi}{4}} \sqrt{\frac{kT}{U_0}} \exp\left(\frac{U_0}{kT}\right)
$$

tK – Kramers time – basis of the Kramers rate theory, postulating that the rate of reactions is limited by diffusion over a high – energy transition rate.

τ= *ɣ/k* - time constant.

Assumption: The energy barrier is high: $U_0 = U(x_0) = 1/2 kx_0^2 >> KT$.

Equation derived in Appendix 4.2.

So far, we have not needed details on the thermal forces that drive Brownian motion and diffusion.

We just needed to assume that the thermal forces were randomly directed

to derive the diffusion equation:

$$
\frac{\partial c(x,t)}{\partial t} = D \frac{\partial^2 c(x,t)}{\partial x^2}
$$

and relate the diffusion coefficient to the drag coefficient:

$$
D = \frac{KT}{\gamma}
$$

$$
p_i = \frac{1}{Z} \exp\left[-\frac{U_i}{kT}\right]
$$

In the presence of force:

$$
\frac{\partial p}{\partial t}(x,t) = D \frac{\partial^2 p}{\partial x^2}(x,t) - \frac{\partial}{\partial x} \left[\frac{F(x)}{\gamma} p(x,t) \right]
$$