

# Calculations in Fourier Space

Rather than doing summations over discrete modes, in many cases we would like to consider a system large enough that a continuum of modes is reasonable. To that end, we return to the string energy and write

$$E = \frac{\gamma}{2} \int_0^L dx h_x^2 \qquad h(x) = \sum_q e^{-iqx} \hat{h}(q)$$

where  $\hat{h}(q)$  is the discrete Fourier transform of  $h(x)$ . Substituting, we have

$$E = \frac{\gamma}{2} \int_0^L dx \sum_q \sum_{q'} (-iq)(-iq') e^{-i(q+q')x} \hat{h}(q) \hat{h}(q')$$

This can be simplified with

$$\int_0^L dx e^{-ipx} = L\delta_{p,0} \rightarrow E = \frac{\gamma L}{2} \sum_q q^2 |\hat{h}(q)|^2$$

We are now able to use the equipartition theorem:

$$\langle |\hat{h}(q)|^2 \rangle = \frac{k_B T}{\gamma L} \frac{1}{q^2}$$

# Calculations in Fourier Space - continued

We denote the thermal average by  $\langle h^2 \rangle$ . A second average of interest is the *system average*, denoted by  $\overline{h^2}$ , defined by

$$\overline{h^2} = \frac{1}{L} \int_0^L dx h^2$$

The thermal average of the system average has a very compact form

$$\overline{\langle h^2 \rangle} = \sum_q \langle |\hat{h}(q)|^2 \rangle$$

This is still a discrete sum. In the continuum limit, we use the fundamental relation

$$\frac{1}{L} \sum_q \rightarrow \int \frac{dq}{2\pi}$$

so

$$\overline{\langle h^2 \rangle} = \frac{k_B T}{2\pi\gamma} \int \frac{dq}{q^2}$$

For this integral we may need to invoke integration cutoffs.

1. Small scale: Molecular lengths  $a$
2. Large scale: Finite system size  $L$

# Calculations in Fourier Space - continued

Accounting for one such cutoff:

$$\overline{\langle h^2 \rangle} = \frac{k_B T}{2\pi\gamma} \left( \frac{1}{q_{\min}} - \frac{1}{q_{\max}} \right) \quad q_{\min} = \pi/L \quad q_{\max} = \pi/a$$

Considering the energy,

$$\langle \overline{h_x^2} \rangle \sim \frac{k_B T}{\gamma} \int \frac{dq}{q^2} q^2 \sim \frac{k_B T}{\gamma} (q_{\max} - q_{\min})$$

As we let  $L \rightarrow \infty$ ,  $q_{\min} \rightarrow 0$ , and as  $a \rightarrow 0$ ,  $q_{\max} \rightarrow \infty$ , so that

$$\langle \overline{h_x^2} \rangle \approx \frac{k_B T}{\gamma} \frac{\pi}{a}$$

# More on Fluctuations

If we generalize this calculation to  $d$  dimensions of space and hence  $d - 1$  dimensions of the surface we find

$$\begin{aligned}\langle \overline{h^2} \rangle &\sim \int \frac{d^{d-1}q}{(2\pi)^{d-1}} \frac{k_B T}{\gamma q^2} \sim \frac{k_B T}{\gamma} \int dq \frac{q^{d-2}}{q^2} \\ &\sim \frac{k_B T}{\gamma} \frac{q^{d-3}}{d-3} \Big|_{q_{\min}}^{q_{\max}} \quad (d > 3) \\ &\sim \frac{k_B T}{\gamma} \ln \left( \frac{q_{\max}}{q_{\min}} \right) \sim \frac{k_B T}{\gamma} \ln \left( \frac{L}{a} \right) \quad (d = 3)\end{aligned}$$

# More on Fluctuations

Now we consider a two-dimensional interface endowed with surface tension and in a gravitational field, with a density difference  $\Delta\rho$  between the fluids on either side. Once again we expand the surface deformation

$$h(\mathbf{r}) = \sum_{\mathbf{q}} e^{-i\mathbf{q}\cdot\mathbf{r}} \hat{h}(\mathbf{q})$$

The quadratic energy functional is

$$\begin{aligned} E &= \frac{1}{2} \int \int dx dy [\sigma(\nabla h)^2 + \Delta\rho g h^2] \\ &= \frac{\sigma A}{2} \sum_{\mathbf{q}} (q^2 + l_c^{-2}) |\hat{h}(\mathbf{q})|^2 \end{aligned}$$

where  $l_c$  is once again the capillary length. Again by equipartition we find

$$\langle |\hat{h}(\mathbf{q})|^2 \rangle = \frac{k_B T}{\sigma A} \frac{1}{l_c^{-2} + q^2}.$$

This shows that the capillary length provides a cutoff on what would otherwise be divergent fluctuation amplitudes as  $q \rightarrow 0$ .

# More on Fluctuations

Introducing both large-scale and small-scale cutoffs, the average variance of the displacement field is

$$\overline{\langle h^2 \rangle} = \frac{k_B T}{2\pi\sigma} \int \frac{q dq}{l_c^{-2} + q^2} \\ \frac{k_B T}{4\pi\sigma} \ln \left[ \frac{1 + 2(\pi l_c/l)^2}{1 + 2(\pi l_c/L)^2} \right]$$

In the thermodynamic limit ( $L \rightarrow \infty$ ), which is now possible at finite  $g$ ,

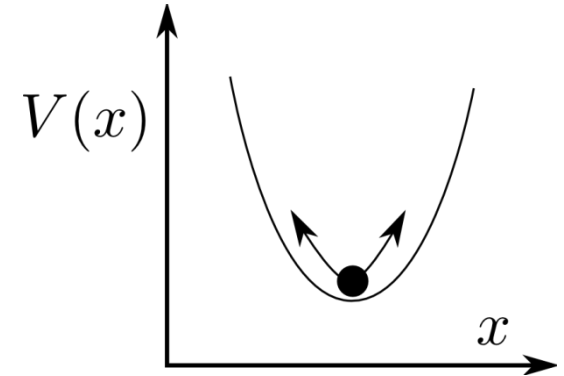
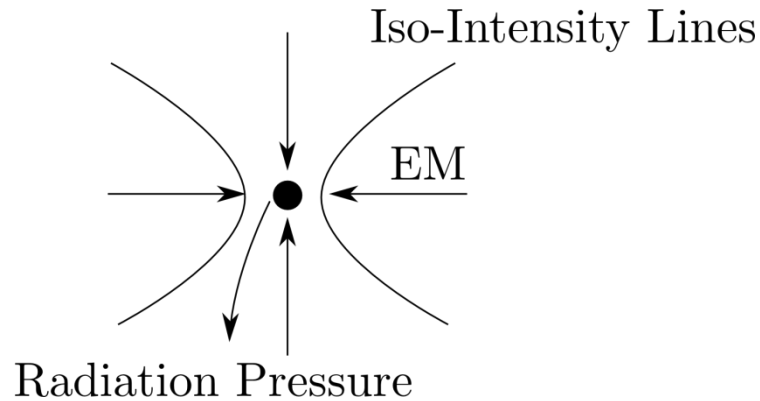
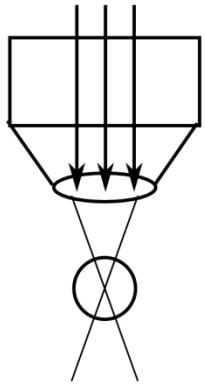
$$\overline{\langle h^2 \rangle} \sim \frac{k_B T}{4\pi\sigma} \ln \left[ 1 + 2 \left( \frac{\pi l_c}{l} \right)^2 \right]$$

It is clear, with  $k_B T \sim 10^{-14}$  erg and  $\sigma \sim 50$  erg/cm<sup>2</sup> that even if  $l_c/l \sim 10^7$  the fluctuations are still on the molecular scale.

# On to Brownian Motion

Brownian motion can be investigated in modern laser trapping systems, first invented in the 1970's at Bell Labs. The focus beam naturally converges on a small diffraction limited region:

Laser Light



Assuming that the trapping potential is quadratic in lateral displacements  $x$ , the overdamped equation of motion of a microsphere in the trap is

$$\zeta \dot{x} = -kx + \eta(t)$$

where  $\eta(t)$  is a random force. For  $\mu\text{m}$  sized spheres and moderate lasers,  $k \sim 10$  fN/nm. For example, attaching spheres onto motor proteins allows the strength of interaction to be determined.

# Analyzing the Langevin Equation

The stall force of motor proteins is a few pN. The relaxation time scale  $\tau$  in the well comes from the spring constant and drag coefficient:  $\tau = \zeta/k \sim 4$  ms.

There are two levels at which we can “solve” the Langevin equation. For any particular realization of the random noise  $\xi(t)$  we can write down  $x(t)$  directly. But we are also interested in *averages over realizations* of the noise, suitable to compare with experimental observations. In the first case, if we rescale the noise term the equation is

$$\dot{x} + \frac{1}{\tau}x = \xi(t)$$

We recognize an integrating factor:

$$e^{t/\tau} \left( \dot{x} + \frac{1}{\tau}x \right) = e^{t/\tau} \xi(t)$$

This allows a direct solution for any particular noise

$$x(t) - x_0 e^{-t/\tau} = \int_0^t dt' e^{-(t-t')/\tau} \xi(t')$$



# Langevin Equation - continued

Now we find averages over realizations of the noise.

$$\langle x(t) - x_0 e^{-t/\tau} \rangle = \int_0^t dt' e^{-(t-t')/\tau} \langle \xi(t') \rangle = 0 .$$

Clearly, the average of the noise must vanish for an unbiased system, so we conclude

$$\langle x(t) \rangle = x_0 e^{-t/\tau} .$$

Now we consider the square of the deviation from simple relaxation:

$$\langle (x(t) - x_0 e^{-t/\tau})^2 \rangle = \int_0^t dt' \int_0^t dt'' e^{-(t-t'')/\tau} e^{-(t-t')/\tau} \langle \xi(t') \xi(t'') \rangle$$

The crucial assumption of the Langevin approach is that the correlation inside the integral is a sharply-peaked function of  $|t - t''|$ , decaying much faster than any relevant timescale of the particle. Calling this function  $\phi(t' - t'')$ , we make the change of variables ( $J = 1/2$ )

$$s = t' + t''$$

$$q = t' - t''$$

# Langevin Equation – continued

The right hand side of the previous equation will then be  
(extending limits to  $\pm\infty$ )

$$\int_0^t dt' \int_0^t dt'' e^{-(t-t'')/\tau} e^{-(t-t')/\tau} \langle \xi(t') \xi(t'') \rangle = \frac{1}{2} e^{-2t/\tau} \int_0^{2t} ds e^{s/\tau} \int_{-\infty}^{\infty} dq \phi(q)$$

where the final integral is just a number ( $\Gamma$ ). The average deviation squared is then

$$\langle (x(t) - x_0 e^{-t/\tau})^2 \rangle = \frac{\Gamma\tau}{2} (1 - e^{-2t/\tau})$$

In the long time limit ( $t/\tau \rightarrow \infty$ )

$$\langle (x(t) - x_0 e^{-t/\tau})^2 \rangle = \langle x(t)^2 \rangle = \frac{\Gamma\tau}{2}$$

However, using the equipartition argument

$$\frac{1}{2} k \langle x^2 \rangle = \frac{1}{2} k_B T$$

$\Rightarrow$

$$\Gamma = \frac{2k_B T}{\zeta}$$

# Langevin Equation - continued

Logically, we assume that there is a  $\delta$  correlation for the noise:

$$\langle \xi(t)\xi(t') \rangle = \frac{2k_B T}{\zeta} \delta(t - t')$$

A further test of the result is to examine the short-time behaviour of the variance in the displacement. If we assume  $x_0 = 0$  and  $t/\tau \ll 1$ , then

$$\langle x^2(t) \rangle \sim \frac{\Gamma\tau}{2} \left( \frac{2t}{\tau} + \dots \right) \sim \frac{2k_B T}{\zeta} t + \dots$$

which is just a random walk in 1D ( $\langle x^2 \rangle = 2Dt$ ). Thus

$$D = \frac{k_B T}{\zeta}$$

and the Stokes-Einstein relation is recovered. The fun calculation is to do this in the presence of inertia (see examples sheet).

# Brownian Diffusion

The diffusion coefficient is just the average of the movement rate per time at long times

$$D = \lim_{t \rightarrow \infty} \frac{1}{6t} \langle (\mathbf{r}(t) - \mathbf{r}(0))^2 \rangle$$

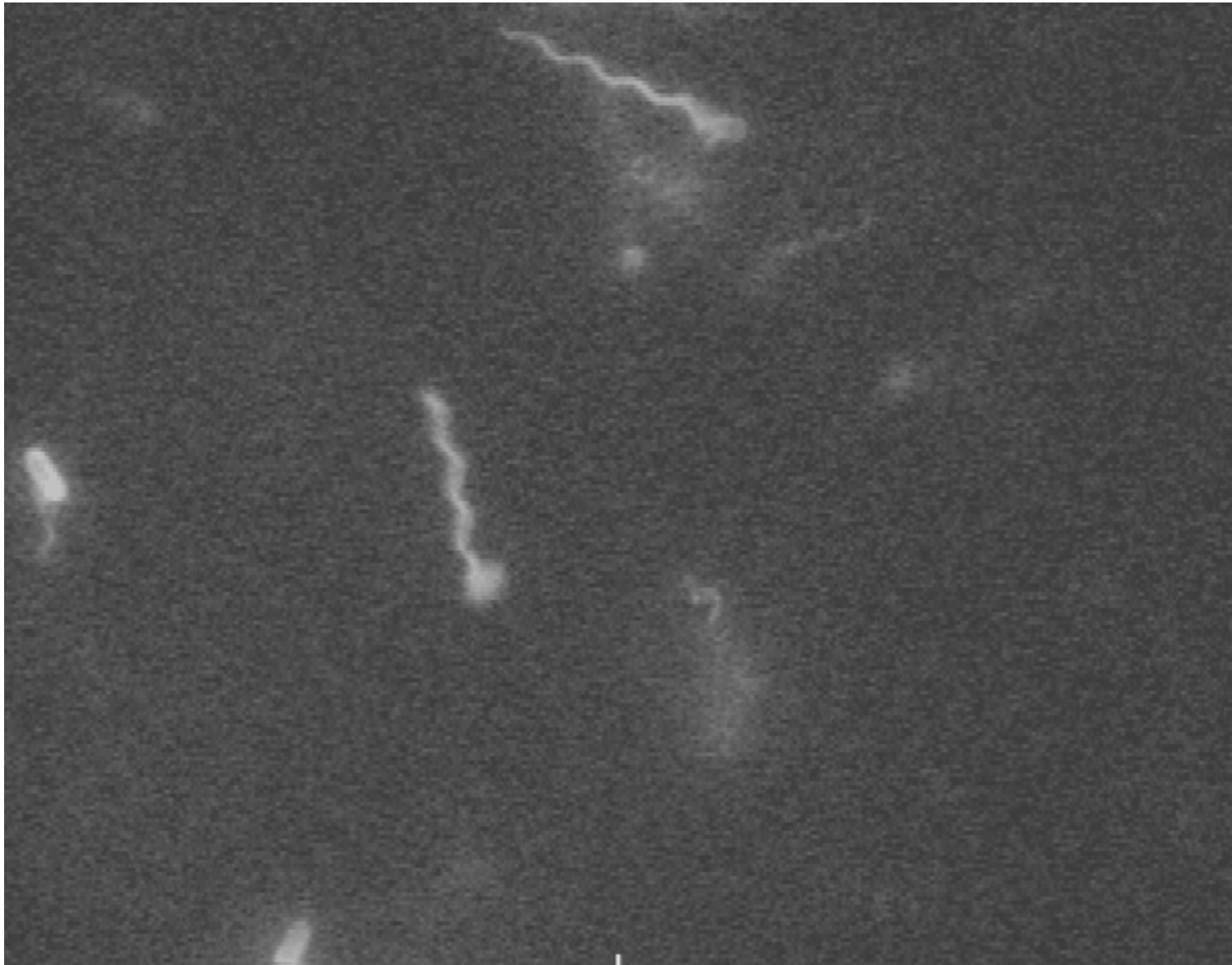
$$\text{where } \mathbf{r}(t) = \mathbf{r}(0) + \int \mathbf{u}(t') dt'$$

The diffusion coefficient term above holds provided that the correlation of velocities ( $\langle \mathbf{u}(t') \cdot \mathbf{u}(t'') \rangle$ ) falls off fast enough. This yields

$$D = \frac{1}{3} \int_0^\infty dt \langle \mathbf{u}(t) \cdot \mathbf{u}(0) \rangle \sim u^2 \tau$$

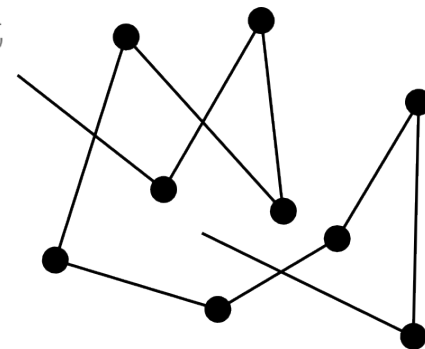
We can apply this to the run-and-tumble locomotion of bacteria. For *E. coli* the average velocity is about  $20 \mu\text{m/s}$ , and the bacteria executes 1s of movement before randomly changing direction. This yields a diffusion coefficient of  $4 \times 10^{-6} \text{cm}^2/\text{s}$ , which is approximately the diffusion coefficient of a small molecule in water.

# Run-and-Tumble Locomotion of *E. coli*



# Brownian Motion and Polymer Statistics

Consider an arbitrary free polymer with each segment labeled as  $\mathbf{r}_n$ . Each segment is followed by another random segment of equal length ( $|\zeta_n| = b$ )



$$\mathbf{r}_{n+1} = \mathbf{r}_n + \zeta_n$$

The end-to-end displacement of the polymer is

$$\mathbf{r}_N - \mathbf{r}_0 = \sum_{n=1}^N \zeta_n \qquad \langle \mathbf{r}_N - \mathbf{r}_0 \rangle = \sum_{n=1}^N \langle \zeta_n \rangle = 0$$

by symmetry. The average of the displacement squared is

$$\langle (\mathbf{r}_N - \mathbf{r}_0)^2 \rangle = \sum_{m=1}^N \sum_{n=1}^N \langle \zeta_m \cdot \zeta_n \rangle = \sum_{m=1}^N \sum_{n=1}^N \delta_{mn} b^2 = Nb^2$$

The similarity with the Langevin formalism is apparent.

# Brownian Motion and Polymers - continued

Let us try to formulate the problem more generally. defining the probability that a polymer will have segment positions at  $\{\mathbf{r}_k\}$  as

$$p = \frac{1}{Z} G(\{\mathbf{r}_k\}) \quad G = e^{-\beta U(\{\mathbf{r}_k\})}$$

Let us suppose that the energy is a sum of near-neighbor interactions plus a contribution from some external potential,

$$U(\{\mathbf{r}_k\}) = \sum_{j=1}^N U_j(\mathbf{r}_{j-1}, \mathbf{r}_j) + W(\{\mathbf{r}_k\})$$

When  $W = 0$ , this is just a random flight model. Either way, this is a local model for the total energy, as it only relies on nearest neighbor interactions.

We then introduce

$$\tau_j(\mathbf{R}_j) = \exp[-\beta U_j(\mathbf{R}_j)] \quad \text{where} \quad \mathbf{R}_j = \mathbf{r}_j - \mathbf{r}_{j-1}$$

and we can take it to be normalized ( $\int d\mathbf{R}_j \tau(\mathbf{R}_j) = 1$ ).

# Brownian Motion and Polymers - continued

We now define the fixed end-to-end-vector partition function as an integral over all degrees of freedom for which the end position is  $\mathbf{R}$  (start at origin):

$$G(\mathbf{R}; N) = \int d\{\mathbf{R}_k\} G(\{\mathbf{R}_k\}) \delta(\mathbf{r}_N - \mathbf{R}) = \int d\{\mathbf{R}_k\} \prod_{j=1}^N \tau(\mathbf{R}_j) \delta\left(\sum_{j=1}^N \mathbf{R}_j - \mathbf{R}\right)$$

We shall see that molecular-level details will be coarse-grained away...

As an example, consider  $\tau$  for a fixed-length segment:

$$\tau(\mathbf{R}_j) = \frac{1}{4\pi\ell^2} \delta(|\mathbf{R}_j| - \ell)$$

Now we use an integral representation of a delta function,

$$\delta\left(\sum_{j=1}^N \mathbf{R}_j - \mathbf{R}\right) = \int \frac{d^3k}{(2\pi)^3} e^{i\mathbf{k} \cdot (\sum \mathbf{R}_j - \mathbf{R})}$$



# Brownian Motion and Polymers - continued

The distribution function is then

$$G = \int \frac{d^3 k}{(2\pi)^3} e^{-i\mathbf{k} \cdot \mathbf{R}} \left[ \int d\mathbf{R}_j \tau(\mathbf{R}_j) \exp(i\mathbf{k} \cdot \mathbf{R}_j) \right]^N$$

The bracketed term is a characteristic function  $K(\mathbf{k}; N)$ , and in this particular case is

$$K(\mathbf{k}; N) = \left( \frac{\sin(k\ell)}{k\ell} \right)^N$$

We expect  $N$  to be on the order of  $R^2$  if dominated by diffusive behavior, and thus quite large. In the limit of large  $N$  (small  $k$ )

$$K(\mathbf{k}; N) \approx \left( 1 - \frac{k^2 \ell^2}{6} + \dots \right)^N \sim \exp(-N k^2 \ell^2 / 6)$$

Inverse Fourier transforming,

$$G(\mathbf{R}; N) = \int \frac{d^3 k}{(2\pi)^3} \exp(-i\mathbf{k} \cdot \mathbf{R}) \exp(-N \ell^2 k^2 / 6) = \left( \frac{3}{2\pi \ell^2 N} \right)^{3/2} \exp\left( \frac{-3R^2}{2N \ell^2} \right)$$