

## Derivations

Momentum of a photon  $\rightarrow P: h\bar{k} = \frac{h}{\lambda_0} = \frac{E}{c} = \frac{hc}{\lambda}$

wavelength in a vacuum  
in a medium,  $\lambda = \frac{\lambda_0}{n}$

Change in momentum:  $|\Delta P| = \frac{2hfn}{c}$

a beam of Power  $P$ , emits  $N$  photons per second:  $N = \frac{P}{hf}$

Force due to photons:  $F = \frac{\Delta P}{\Delta t} = \frac{2hfnP}{hf c} = \frac{2nP}{c}$

$\Rightarrow$  If surface is not a perfect reflector:  $F = \frac{2nP}{c} \times \Gamma$  (up reflection coefficient)

To account for a spherical source:  $|F| = \frac{\Omega n P}{c}$  ( $\Omega$  = trap efficiency)

Mie regime:

$$2r \ll \frac{\lambda}{\pi}$$

Rayleigh regime

$$2r \gg \frac{\lambda}{\pi}$$

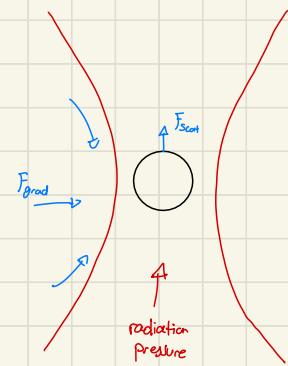


$$|F_{\text{scat}}| = \frac{I_0 \sigma n_m}{c}$$

Scattering Cross Section  $\rightarrow \sigma = \frac{12.8 \pi^5 r^6}{3 \lambda^4} \left( \frac{n^2 - 1}{n^2 + 1} \right)^2$   $n_m: \frac{n_p}{n_m}$

$$|F_{\text{grad}}| = \frac{2\pi\sigma}{c n_m^2} \nabla I_0$$

Fibrillability  $\rightarrow \sigma = n_m^2 r^3 \left( \frac{n^2 - 1}{n^2 + 1} \right)^2$



Most particles used fall in between the regimes so Eq. 7 is used and  $\Omega$  accounts for uncertainties.

## Trapping Force:

Treating the trap as a Hookean Spring  $\rightarrow$  particle is an over-damped oscillator:

$$m\ddot{x}(t) + \gamma\dot{x}(t) + Kx(t) = F_b(t)$$

↓  
 Particle mass  
 ↓  
 Viscous drag coefficient of medium  
 ↓  
 optical trap stiffness  
 ↓  
 K force due to thermal fluctuations

$m\ddot{x}(t)$  is the inertial force. Trapping is in low Reynolds number regime,  $Re \ll 1$ , viscous forces dominate.  $\therefore m\ddot{x}(t)$  is neglected

### Stokes Drag Method

Stokes law

$$F = 6\pi r \eta$$

Stokes viscous force:

$$F_s = 6\pi r \eta \dot{x}(t)$$

Trapping, if the thermal fluctuations force is small (assume = 0):

$$-Kx(t) = \gamma\dot{x}(t)$$

↓  
 F<sub>T</sub>  
 Trapping force

Stokes force



- More contains at increasing velocities until  $F_s > F_T \rightarrow$  particle leaves trap
- Velocity at which the particle leaves the trap  $\rightarrow F_s = \gamma\dot{x}(t)$

$$\frac{1}{F_T}$$

## ACF and MSD

Overdamped Langevin eqn:

$$m\ddot{x}(t) + \gamma\dot{x}(t) + Kx(t) = F_b(t)$$

Solving for  $x(t)$  gives the autocorrelation function (ACF):

$$\langle x(t+\tau) x(t) \rangle = \langle x^2 \rangle e^{-\tau/\tau_c}$$

↗ auto correlation time  
 ↗ decay time

$$\tau_c = \frac{\gamma}{K} \rightarrow 6\pi r \eta$$

$$\vdots$$

$$\eta = \frac{\tau_c k_B T}{6\pi r \eta \tau_c}$$

MSD of  $x(t)$ :

$$\langle \Delta x^2(\tau) \rangle = \langle (x(t+\tau) - x(t))^2 \rangle$$

related to ACF by:

$$\langle \Delta x^2(\tau) \rangle = 2\langle x^2 \rangle - 2\langle x(t+\tau) x(t) \rangle$$

$$\therefore \langle \Delta x^2(\tau) \rangle = \frac{2k_B T}{\eta} \left[ 1 - e^{-\frac{\tau}{\tau_c}} \right]$$

So ACF and MSD are complementary

### Equipartition method

so the optical trap is 1D

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

↳ positional variance

in a potential  $U(x)$

Probability of finding the particle at  $x$  is given by a Boltzmann distribution:

$$P(x) \propto \exp\left(-\frac{U(x)}{k_B T}\right) \propto H(x)$$

↳ Histogram of particle position

$$U(x) = \frac{1}{2} k (x - x_0)^2$$

$$H(x) \propto P(x) = C \exp\left(-\frac{k}{2k_B T} (x - x_0)^2\right)$$

gaussian



• measure position ( $x$ )  $\rightarrow$  calculate  $H(x)$

• fit Gaussian to determine  $\langle x^2 \rangle$  then  $\eta \Rightarrow$

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

in both  $x$  and  $y$

goal

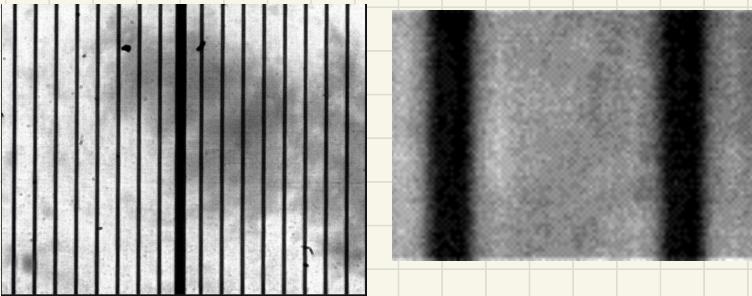
?

## Day 2:

### Calibration

Started with the 2  $\mu\text{m}$  silica beads. We concluded that the lens was dirty, therefore we cleaned it with methanol. The light was also dirty, this was cleaned to the best of our abilities but still shows in photos.

We calibrated the magnification of the microscope.



This is a graticule seen under the microscope.

Each increment is 10  $\mu\text{m}$ .

Measured the graticule in both  $x$  and  $y$  to account for

Any camera distortion. However, both directions measured the

Same:

$$10 \mu\text{m} = (75 \pm 2.5) \text{ px} \quad \text{in } x$$

$$10 \mu\text{m} = (75 \pm 2.5) \text{ px} \quad \text{in } y$$

$$\alpha_f = f \frac{\alpha_x}{51}$$

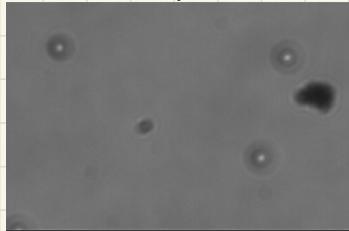
$$1 \text{ px} = 0.133 \mu\text{m} \pm 0.005 \mu\text{m}$$

## Brownian motion

Day 2-3

Prepared a sample of diluted  $2\text{ }\mu\text{m} \pm 5\%$  microbeads.

Tracked the motion of particles. Made an effort to maximize fps and minimize lost frames. Also tried to keep ROI bright



Dried a particle on a slide

The track of a stationary particle was measured to account for thermal vibrations

A mean-squared displacement was plotted for this stationary particle.

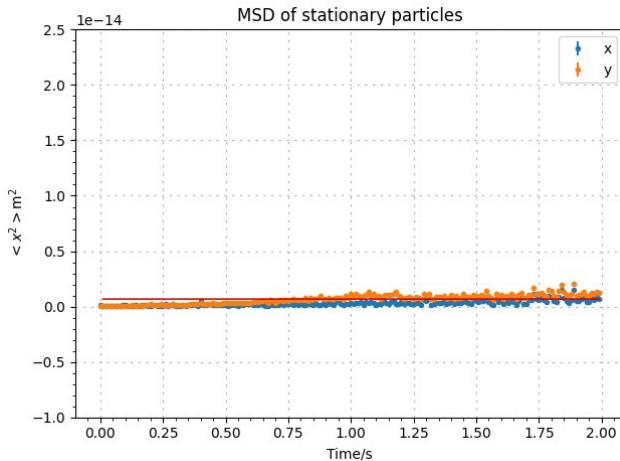
An average value was taken of this graph. This average was subtracted from all subsequent data, thus accounted for the

Systematic error.

However, this error was between 2 and 3

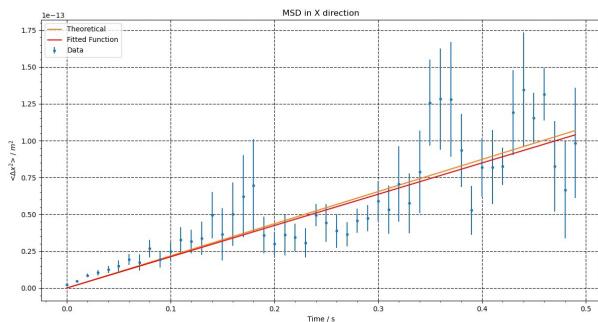
orders of magnitude smaller than data so

was neglected.



Then plotted MSD for a series of particles. An example plot shown below

Mean Squared Displacement (MSD) graphs



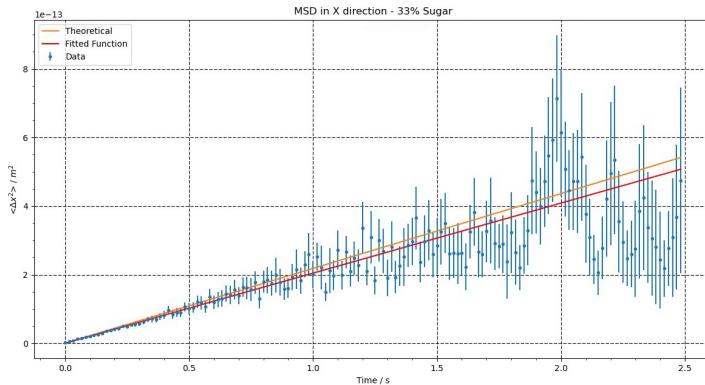
This is a small part of the plot.

The large data points have large error due to the nature of the plot.

Therefore, the first few data points were taken.

This part of the plot is linear.

To refine this plot, we maximized the frame rate of the video. This meant more data points could be taken in the linear region.



The equation for the MSD is

$$\langle \Delta r^2 \rangle = 2nDt$$

where  $n$  is the number of dimensions

$D$  is the diffusion constant.

$$\langle \Delta x^2 \rangle = 2D\tau$$

Therefore the gradient of the MSD is:  $m = 2D$

$$D = \frac{k_B T}{6\pi\eta r}$$

So the viscosity of the medium is:

$$\eta = \frac{k_B T}{6\pi D r} = \frac{k_B T}{3\pi r} \quad T = 298.0 \pm 0.5 \quad r = 2.06 \mu\text{m} \pm 5\%$$

Error on the gradient found from polyfit:

$$\sigma_\eta = \eta \sqrt{\left(\frac{\sigma_T}{T}\right)^2 + \left(\frac{\sigma_r}{r}\right)^2 + \left(\frac{\sigma_m}{m}\right)^2}$$

The values of Viscosity Calculated are as follows:

Viscosity / mPaS	Viscosity Uncertainty / mPaS
0.611	0.083
0.660	0.049
1.003	0.144
0.836	0.057
0.903	0.075
0.905	0.091
0.818	0.103
1.063	0.133
0.911	0.125
0.610	0.055
0.813	0.079
0.570	0.038
0.301	0.029
0.906	0.139
1.097	0.140
1.302	0.182

This gives a viscosity

$$\eta = 0.882 \pm 0.104 \text{ mPaS}$$

However if we calculate the std

$$\sigma = 0.242$$

and remove any data points further

than 2 $\sigma$  away from the mean

Then the table becomes:

Viscosity / mPaS	Viscosity Uncertainty / mPaS
0.611	0.083
0.660	0.049
1.003	0.144
0.836	0.057
0.903	0.075
0.905	0.091
0.818	0.103
1.063	0.133
0.911	0.125
0.610	0.055
0.813	0.079
0.570	0.038
0.301	0.029
0.906	0.139
1.097	0.140
1.302	0.182

This gives a

viscosity of

$$\eta = 0.867 \pm 0.10 \text{ mPaS}$$

which is consistent with

the viscosity of water

at  $T = 298 \text{ K}$

$$\eta_{\text{true}} = 0.958 \text{ mPaS}$$

## Optical trap Stiffness - Equipartition method

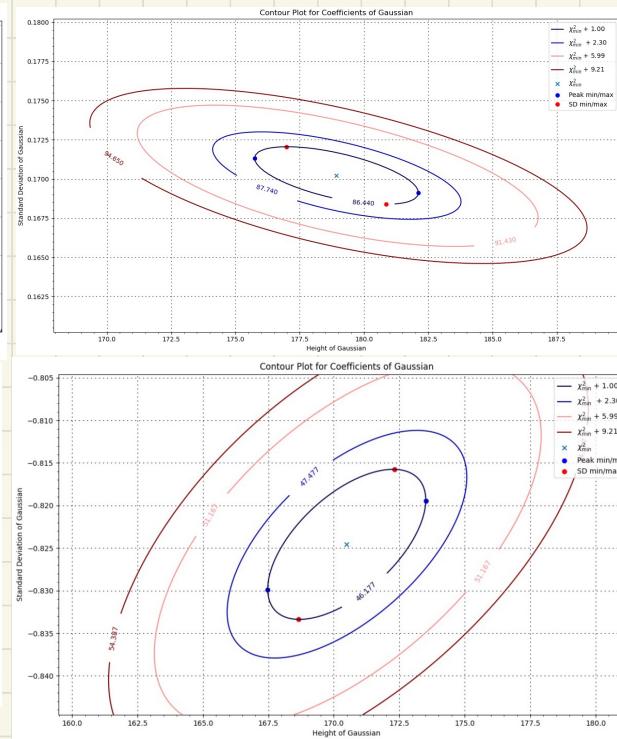
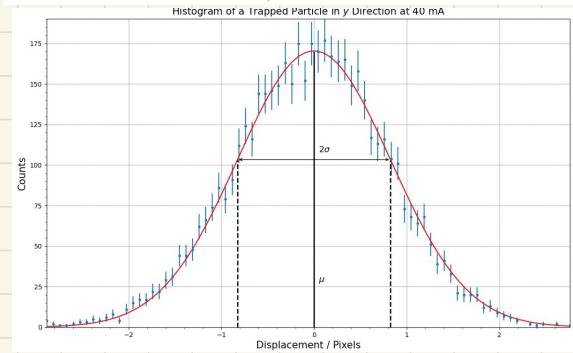
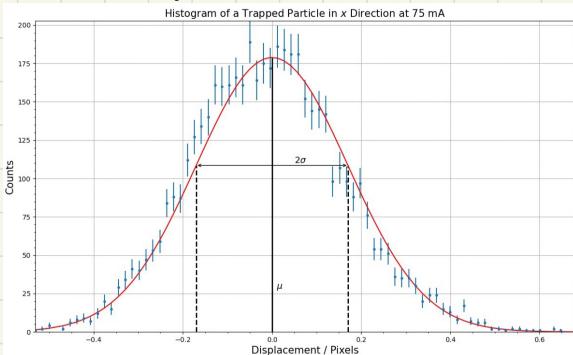
day 4 - 5

Trapped particles using the laser.

Tracked the particles movement using blinder.

Took the mean of all position in x and y and took this away from all data points to normalise the data around  $(\bar{x}, \bar{y})$ .

Plotted the position of the particle in a histogram as we expect this to be gaussian. Example shown below:



As seen above, we fitted a Gaussian to the data. We did this by taking the equation for a Gaussian  $G(x) = A \exp\left(\frac{(x - \mu)^2}{2\sigma^2}\right)$

and varying A and C whilst minimising the chi square. A is the peak and C is the standard deviation.

We also tested the Kurtosis and Skewness of the data to evaluate whether a gaussian could be fitted.

Kurtosis  $\approx 0$  for all gaussians and Skewness also  $\approx 0$

$$\text{Kurtosis} = n \frac{\sum (x_i - \bar{x})^4}{(\sum (x_i - \bar{x})^2)^2} \quad ; \quad \text{Skewness} = \frac{1}{n\sigma} \sum (x_i - \bar{x})^3$$

Data is all mesokurtic with no skew so all suitable for Gaussians

The potential caused by the laser has a 'harmonic' form :  $U(x) = \frac{1}{2} k(x - x_0)^2$

and  $P(x)$  or to the particle's position (Histogram)

$$H(x) = A \exp\left(-\frac{k}{2kT} (x - x_0)^2\right) \quad x_0 = 0 \text{ because we centered at 0.}$$

$$\text{Comparing this to the Gaussian we fit: } \frac{B}{kT} = \frac{1}{C^2}$$

$$B = \frac{kT}{C^2}$$

Where B is the optical trap stiffness and C is the standard deviation

Particle's position in a potential well is given

by a Boltzmann distribution

$$P(x) \propto \exp\left(-\frac{U(x)}{kT}\right)$$

After fitting a gaussian we varied the Coefficients ( $A$  and  $c$ ) and calculated the  $\chi^2$  for different values.

From this we plotted Contour plots.

The Contour Corresponding to  $\chi^2 = \chi^2_{\text{min}} + 1$  has max and min Coefficients values that are a Standard deviation away from the value that give  $\chi^2_{\text{min}}$ .

Therefore finding the distance between the max and min values of each Coefficient on this Contour Corresponds to 2 or on that coefficient. Half of this is the associated error.

This method gives an error associated with  $C$ .

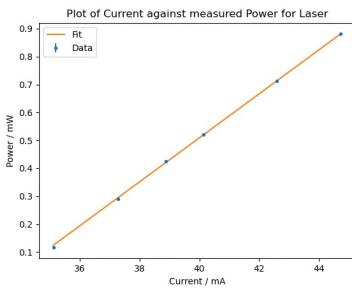
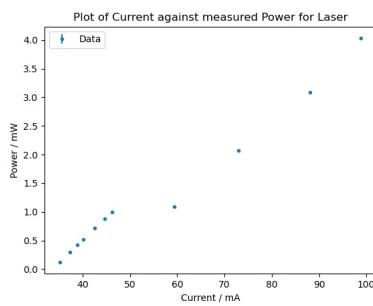
Current / mA	Kappa / pN microm^-1	Kappa Error / pN microm^-1
35	0.083	0.004
40	0.302	0.017
45	0.462	0.026
50	0.710	0.039
55	1.069	0.055
60	1.382	0.075
65	1.411	0.075
70	1.595	0.087
75	1.947	0.091
80	2.713	0.143
85	2.514	0.134
99	2.631	0.169

$$\text{Now using } k = \frac{keT}{C^2} \text{ with } T = 298.5 \pm 0.5$$

$$\text{So } \sigma_k = k \sqrt{\left(\frac{\sigma_T}{T}\right)^2 + \left(2 \frac{\sigma_C}{C}\right)^2}$$

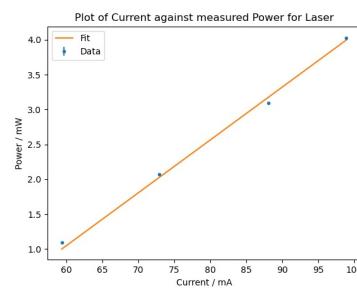
the optical trap stiffness,  $k$ , was calculated with an associated error for a range of currents supplied to the laser.

As expected,  $k$  increased ( $C$  decreased) as  $I$  increased. This is because the laser is weaker. So the particle can move more from the equilibrium position.



$$y = 0.079x - 2.642$$

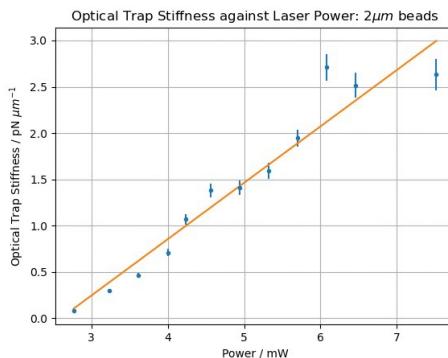
As the intercept should be 0 we take the



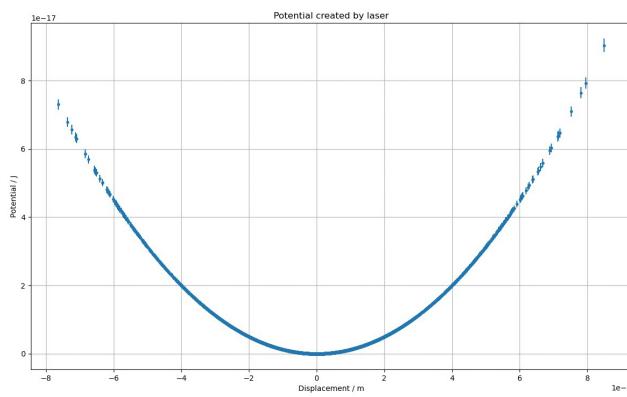
$$y = 0.076x - 3.500$$

relationship to be  $P = 0.0775I$

Now finally, optical trap stiffness was plotted against laser power.



Also, despite doing no analysis on it, we plotted the potential created by the laser by assuming a harmonic potential:



$$\chi^2_{\text{red}} = 10.94$$

$$y = (6.608 \pm 0.003) x - (1.579 \pm 0.069)$$

This follows a linear relationship as expected. Trap stiffness increases linearly with laser power.

Although the relationship depends on the laser parameters, the values of  $k$  were of the same order as values from literature.

The  $\chi^2$  is too big, this is because errors were underestimated or some of the data points didn't fit the data. Taking more repeats or omitting data points would help with this.

$$U = \frac{1}{2} k (x - x_0)^2$$

$$x_0 = 0$$

$$\bar{U}_V = U \frac{\partial k}{k}$$

## Optical Trap Stiffness - Stokes method

day (6)

By accelerating a trapped particle through the medium the drag force increases as  $F_d = \gamma \dot{x}$ .

Therefore, the particle will reach a point where  $F_s > F_d$  and the particle leaves the trap. The velocity at which this occurs can be used to calculate the optical trap stiffness.

$$F_T = F_s$$

$$k \dot{x} = \gamma \dot{x}$$

$$k = \frac{\gamma \dot{x}}{\dot{x}}$$

$$\text{where } \gamma = 6\pi r n$$

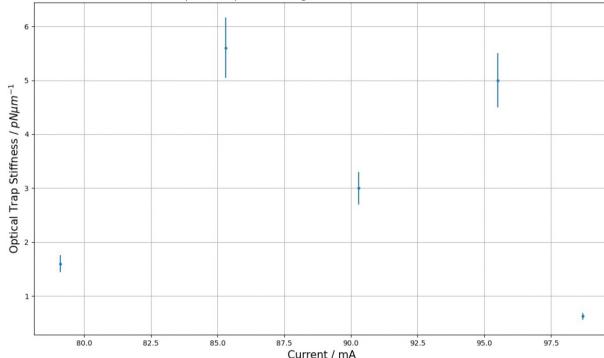
and  $\dot{x}$  is the displacement

from the equilibrium point.

$$k = \sqrt{\left(\frac{\sigma_r}{r}\right)^2 + \left(\frac{\sigma_z}{z}\right)^2 + \left(\frac{\sigma_x}{x}\right)^2}$$

We repeated this for a range of low powers; however, there was no correlation between the data so this was not pursued.

Optical Trap Stiffness against Current - Stokes Method



One reason this didn't work is it is not possible to see all particles in the different planes of the sample. However, the trapped particle can still collide with this.

Despite taking repeats it appears this was happening more than we thought.

A more dilute sample would help to combat this.

## Optical trap Stiffness - MSD

(day 6)

The equation of an MSD of a particle in an optical trap with stiffness  $k$  is given by:

$$\langle \Delta x^2(t) \rangle = \frac{2kT}{k} \left[ 1 - \exp\left(-\frac{kT}{\gamma}\right) \right]$$

Taking the log of both sides:

$$\begin{aligned} \log(\Delta x^2) &= \log\left(\frac{2kT}{k}\left(1 - \exp\left(-\frac{kT}{\gamma}\right)\right)\right) \\ \log(\Delta x^2) &= \log\left(\frac{2kT}{k}\right) + \underbrace{\log\left(1 - \exp\left(-\frac{kT}{\gamma}\right)\right)}_{\substack{k \approx 10^6 \\ \gamma \approx 10^3 \\ \text{so } e^{-x} \approx 1 - x \text{ for small } x.}} \\ \log(\Delta x^2) &\approx \log\left(\frac{2kT}{k}\right) + \frac{kT}{\gamma} \quad \text{for small } x. \end{aligned}$$

Although we are only interested in the intercept.

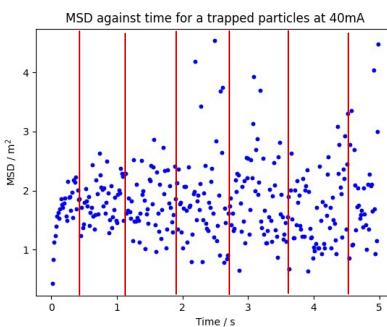
When plotting the raw data, the data was very noisy. So we divided it into bins.

In each bin we took a mean and SD. We removed any data points more than 2 SD away from the mean.

A new mean was then found in each bin, the error on these data points (BH plot) were

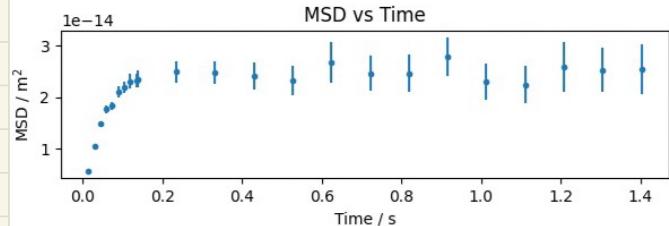
$$\sigma = \sqrt{\frac{\sum(\sigma_i)^2}{N}}$$

Before Binning



After Binning

MSD analysis in y direction at 40.0mA



We then took the log:

From  $\log(\Delta x^2)$ :

$$\sigma_{\log(\Delta x^2)} = \sqrt{\left(\frac{df}{dz}\right)^2 \sigma_z^2}$$

$$f = \ln z$$

$$= \frac{df}{dz} \sigma_z$$

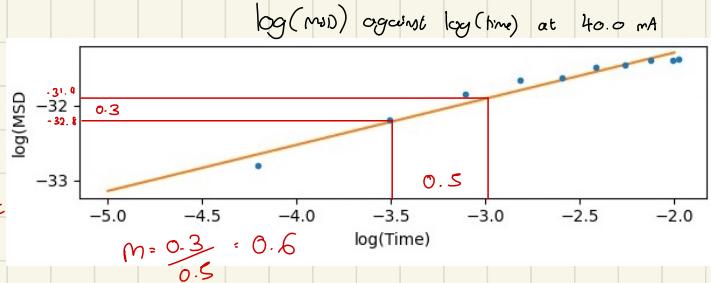
$$= \frac{1}{z} \sigma_z$$

$$\sigma_{\log(\Delta x^2)} = \frac{\sigma_{\log z}}{\Delta x^2}$$

$$y = 0.6x + C$$

$$-32.8 = 0.6(-3.5) + C$$

$$-30.7 = C$$



We then used polyfit to fit a linear function to this data before the cornering frequency.

The y intercept is given by:

$$C = \ln\left(\frac{2k_B T}{h}\right)$$

$$e^C = \frac{2k_B T}{h}$$

$$B = \frac{2k_B T}{e^C}$$

$\sigma_C$  comes from the covariance matrix of the fit

$$\begin{pmatrix} \sigma_m^2 & \sigma_{mc} \\ \sigma_{mc} & \sigma_c^2 \end{pmatrix}$$

$$\sigma_f = \frac{ds}{dx} \sigma_x$$

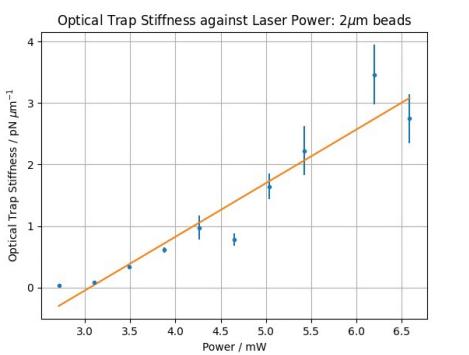
$$= -f \sigma_x \quad f = e^{-Cx}$$

$$\underline{\sigma_B = k \sigma_C}$$

So we calculated it and its error for a range of laser power and plotted as a straight line. Repeated measurements at the same current so

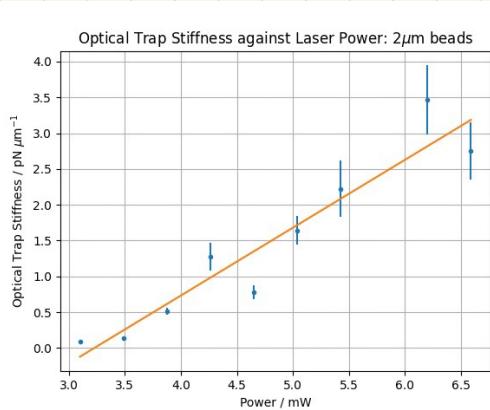
$$\text{used } \sigma_N = \sqrt{\frac{\sum(\sigma_i)^2}{N}}$$

for error on the mean.



The errors on the first data points are small so they inflate the  $\chi^2$ .

Removing the first data point as an anomaly to the linear fit improved the  $\chi^2_{\text{red}}$  to 49.6



$$y = (0.95 \pm 0.12)x - (3.06 \pm 0.58)$$

This method does not produce a line equation that is consistent with the equipartition method.

The ep method has  $\chi^2_{\text{red}} \approx 10$  whereas the log(mw) method has  $\chi^2_{\text{red}} \approx 50$ . This means the ep method is more suitable for a linear fit.

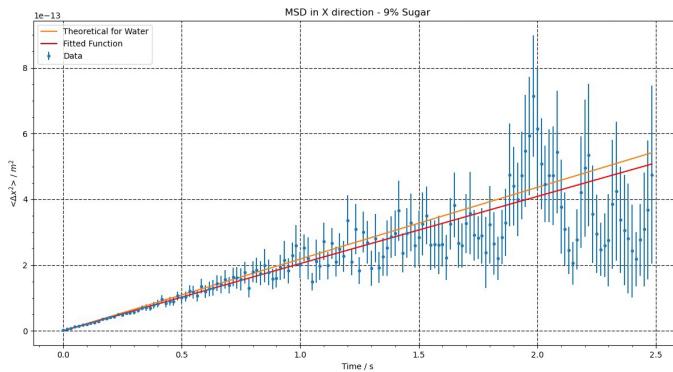
Also the errors on B are smaller in ep. Therefore, the equipartition method is more suitable for calculating B.

# Varying Concentration of Sugar

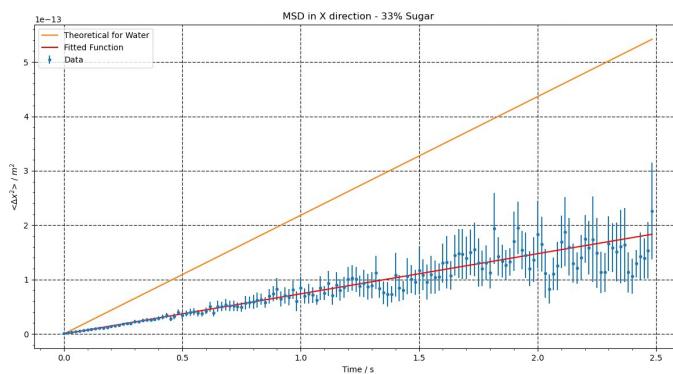
day 7

As an extension we tracked the Brownian motion of particles in solutions of varying sugar concentration.

For this we followed the same method as before to calculate the viscosity.



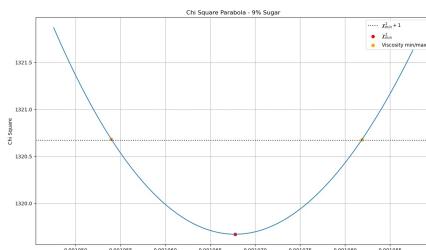
Example of an MSD for 9% Sugar



Example of an MSD for 33% Sugar.

The orange line is has a gradient corresponding to  $\eta = 0.001 \text{ Pas}$  (water). As  $m = \frac{2k_B T}{6\pi r_n}$  then we expect that solutions with higher viscosity than water have a shallower gradient.

After fitting a straight line to the MSD we performed a 1 dimensional varying chi square plot as shown below.



We varied the viscosity around the value that gave the minimum chi square.

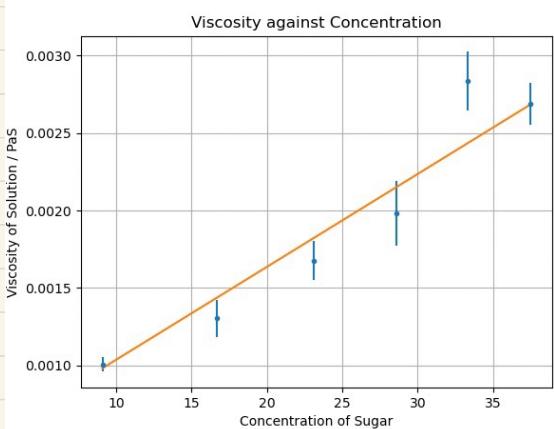
The chi square was calculated for each of these viscosities.

These were plotted against each other.

The values of viscosity that have a  $\chi^2 = \chi^2_{\text{min}} + 1$  are 1 $\sigma$  away from the calculated viscosity.

This is therefore the error on viscosity.

So after finding errors for each viscosity, viscosity was plotted as a function of Sugar Concentration



This data appeared linear to us so we fitted a linear function to it  
This gave a chi-square of  $\chi^2_{\text{red}} = 1.31$   
and

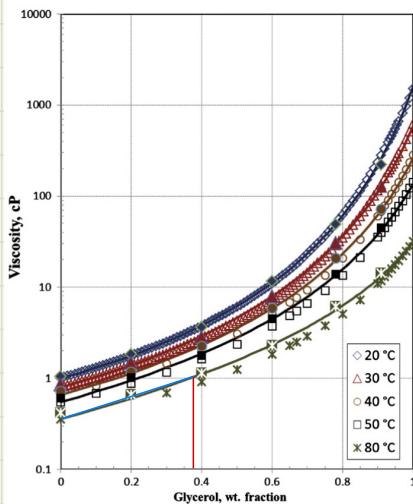
$$y = (5.99 \times 10^{-5} \pm 3.24 \times 10^{-6})x + (4.37 \times 10^{-4} \pm 1.04 \times 10^{-4})$$

The maximum concentration of sugar that we could measure Brownian motion for was 37.5%.

Anything above this and we found that the viscosity was too high for Brownian motion for 2μm beads.

Note that if we had used smaller beads we could have tested higher concentrations.

In theory the Viscosity is not linear with Concentration it follows a somewhat exponential relationship.



However, we only tested up to a concentration of 37.5%  
Looking at the graph the relationship is linear up to this concentration  
So our data matches theory.

"Physical Properties of aqueous glycerol Solutions"

Takamura, Fischer, Morrow, Journal of Petroleum Science and Engineering (98-99) 2012.

# Simulating Brownian motion

day 8...

Due to limitations of the equipment ( limited particle sizes, fixed temperature, can only track in 2 dimensions)

We wanted to simulate Brownian motion to investigate these variables.

To start we consider the Langevin equation:

$$m \frac{d^2x}{dt^2} = -\gamma \frac{dx}{dt} + b \propto + \sqrt{2k_B T \gamma} W(t)$$

$\sqrt{2k_B T \gamma} W(t)$  is the force arising from random impulses (Brownian motion)

For a free particle  $b=0$  so

$$m \frac{d^2x}{dt^2} = -\gamma \frac{dx}{dt} + \sqrt{2k_B T \gamma} W(t)$$

This is a 'white noise' term making the ODE stochastic

White noise is discontinuous and has infinite variation

Using finite difference:

$$m \left[ \frac{x_i - 2x_{i-1} + x_{i-2}}{\Delta t^2} \right] = -\gamma \left[ \frac{x_i - x_{i-1}}{\Delta t} \right] + \sqrt{2k_B T \gamma} W(t)$$

Properties of  $W(t)$ :  $\langle W(t) \rangle = 0$   $\langle W(t)^2 \rangle = 1$   $W(t_1)$  independent of  $W(t_2)$ . Cannot approximate with instantaneous values at  $t_i$

It is convenient and valid to ignore the inertial term ( $m=1$ )

(replace  $W(t)$  with  $\frac{w_i}{\sqrt{\Delta t}}$ )

$$\gamma \dot{x} = \sqrt{2k_B T \gamma} W(t)$$

$$\dot{x} \sim \sqrt{2D} W(t)$$

where  $w_i$  are normally distributed

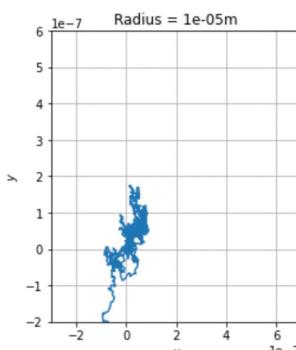
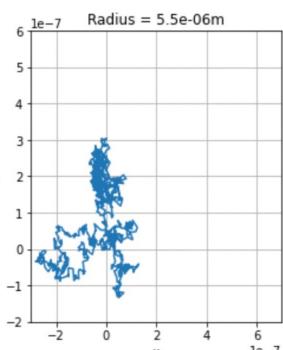
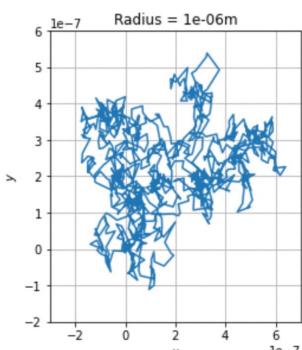
$$\frac{x_i - x_{i-1}}{\Delta t} = \sqrt{2D} \frac{w_i}{\sqrt{\Delta t}}$$

random numbers with unit variance

$$x_i = x_{i-1} + \sqrt{2D \Delta t} w_i$$

This equation was then used to simulate Brownian motion for a range of particle sizes

## Showing Brownian motion for varying particle sizes

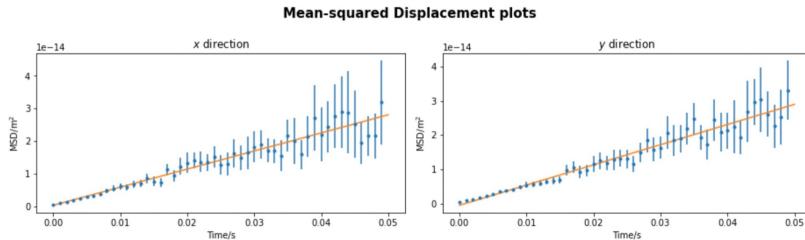


It can be seen from these plots that a particle of larger radius takes larger steps than a particle of smaller radius.

This is because the drag force is  $F_d = 6\pi r \eta v$ , larger particles experience a larger drag so move less far after a collision.

These Brownian motion paths were then used to plot on MD using the method used originally.

An example of this is shown below.



The gradient of this was used to calculate the viscosity (used as a check for code)

This returned a viscosity  $\approx$  to the input viscosity.

## Simulating optical trap

In an optical trap the Langevin equation becomes

$$\dot{\underline{r}}(t) = -\frac{1}{\delta} \underline{b} \cdot \underline{r}(t) + \sqrt{2D} \underline{W}(t) \quad \text{where } \underline{b} = \begin{pmatrix} b_x \\ b_y \\ b_z \end{pmatrix}$$

$$b_x = b_y \neq b_z$$

due to scattering forces

and gravity in the  $z$  direction

and a different intensity

profile.

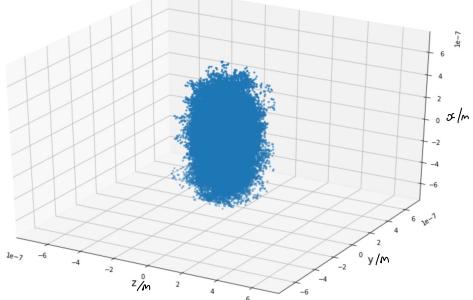
Using the same method to solve this:

$$\underline{r}_i = \underline{r}_{i-1} - \frac{1}{\delta} \underline{b} \cdot \underline{r}_{i-1} \Delta t + \sqrt{2D \Delta t} \underline{w}_i$$

This was then used to simulate particles in an optical trap in 3 dimensions

for  $b_x = b_y = 10^{-6}$  and  $b_z = 10^{-7}$  and  $R = 2 \mu\text{m}$ .

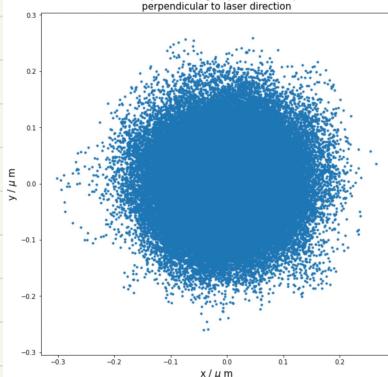
Motion of a Particle in an Optical Trap



The particle has more variation in the  $z$  direction as expected.

In the  $x$ - $y$  plane (perpendicular to the propagation of the laser):

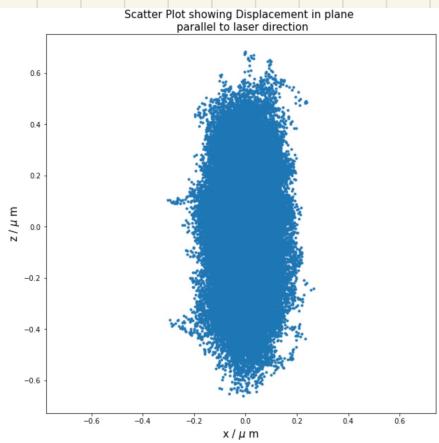
Scatter plot showing displacement in plane perpendicular to laser direction



This is symmetric as expected,  $b_x = b_y$ .

The particle has a higher probability of being in the center when  $r$  increases.

In the  $\alpha$ - $z$  plane:

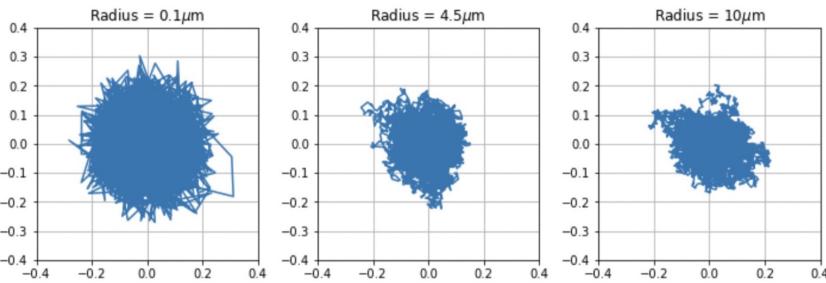


Again, as  $b_z \ll b_x$  the particle is more free to move in the  $z$  direction. This is shown by a higher variation in position in the  $z$  direction.

## Varying $R$ in an optical trap

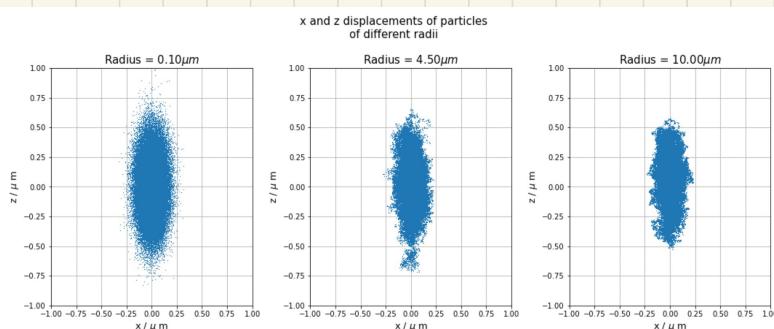
We repeated the process above for 3 different radii of beads to investigate how different sized beads behave in optical trap.

X and Y displacements of particles of different radii



The particles' variation decreases as radius increases. This makes intuitive sense as the step size is smaller (from before). So remain closer to the equilibrium position. This is because a smaller particle is subject to a smaller drag force ( $F = 6\pi r N V$ ). So will move further for a given impulse.

In the  $\alpha$ - $z$  plane:

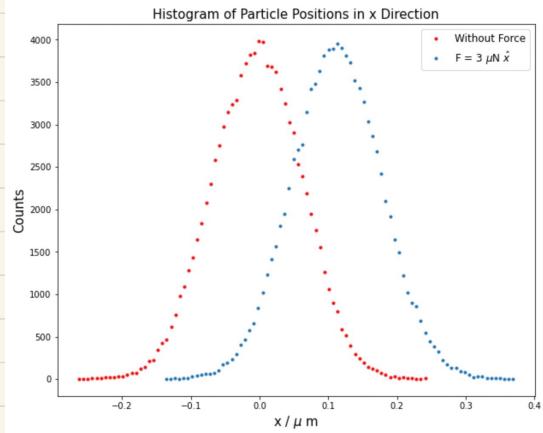


## Adding a Constant force

Adding a Constant force to the System will Shift the equilibrium position of the particle.

If we plot a histogram of the particle's position we expect a gaussian.

By applying a Constant force the mean of this gaussian will shift.



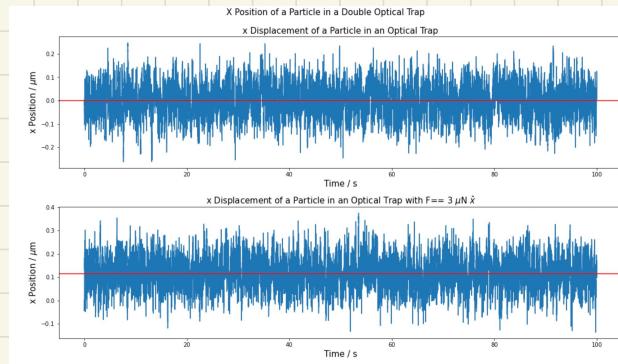
As expected the equilibrium position shifted by about  $0.11 \mu m$  with a force of  $F = 3 \mu N \hat{x}$

The total force in the  $x$  direction has is

$$F = (-k_x x + 3 \mu N) \hat{x}$$

The motion is unchanged in the  $y$  and  $z$  direction.

Another way of visualising this:



Again, it can be seen that the equilibrium position has shifted by  $0.11 \mu m$ .

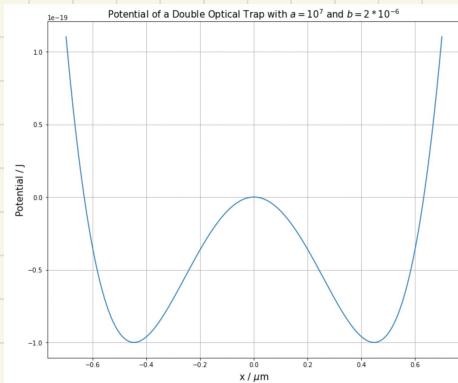
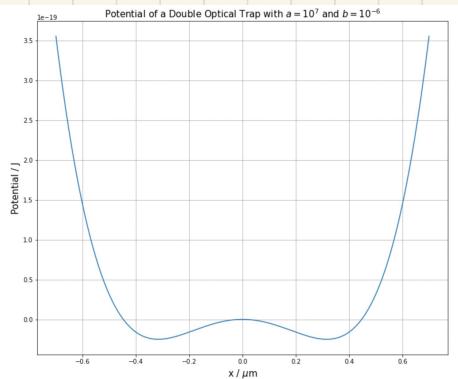
# Double optical trap

$$x_i = x_{i-1} + \frac{1}{\delta} \left[ -ax_{i-1}^3 + bx_{i-1} \right] \delta t + \sqrt{2D\delta t} \omega_i$$

Here we study the effect of a single particle in a double optical trap:

The potential caused by a double optical trap is of the form  $U(x) = \frac{ax^4}{4} - \frac{bx^2}{2}$

Graphing this:



This means the force of the optical trap in  $x$  is  $F = -ax^3 + bx$

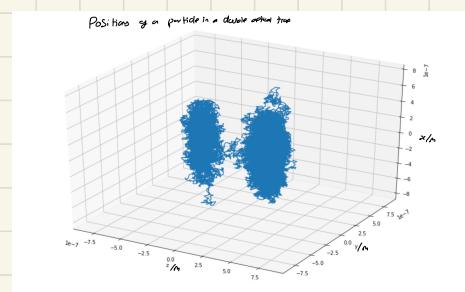
Just by looking at this it is clear to see that there are 2 wells that the particle can sit in.

If the barrier between the wells is low enough that random Brownian motion fluctuations can push the particle into another well

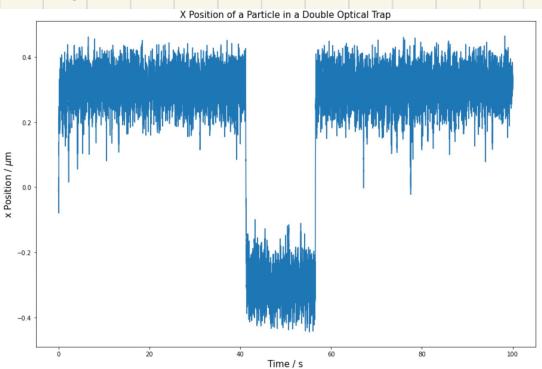
Here  $b$  is such that the size of the barrier is larger so the chance of a particle switching wells is unlikely.

By refining the values of  $a$  and  $b$  so the particle can jump between the wells (Kramers transitions)

we produced this plot of particle positions:

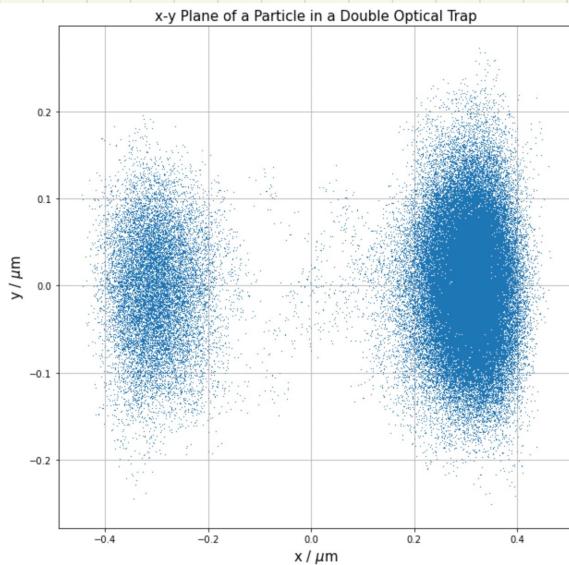


Plotting only the  $\infty$  position:



This makes it clear when the particle jumps between traps.

And Plotting the  $\infty$ -y plane



### Note

- Now two equilibrium positions
- No Hooke's law in  $\infty$  direction
- Force in  $\infty$  is stronger than with 1 trap
- More spread in y.

## Viscosity Gradient

Here we consider when the viscosity of the medium is not constant.

We use the viscosity gradient:

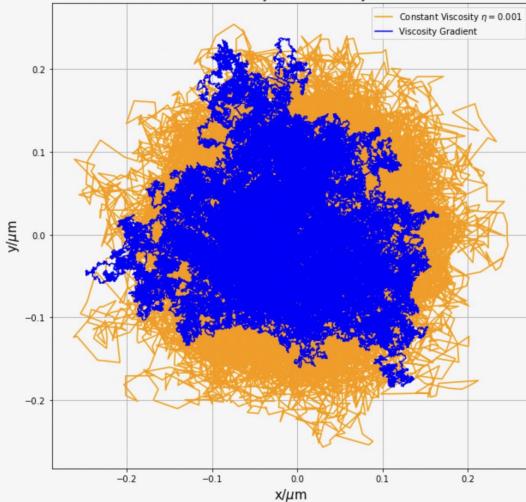
$$\gamma = \gamma_0 + 4 \times 10^8 (x + y) \quad \text{where } \gamma_0 = 0.001 \text{ Pas (water)}$$

This means that as the particle gets further from the origin the viscosity of the medium increases.

This means that the step size of the particle decreases as it gets further from the origin.

This, in turn, means the particle should be confined to a smaller area for a constant  $k$ .

Paths taken by Particles in x-y Plane



This matches all expectation.

More landscapes could have been explored such as the trap being on an interface of two media with different viscosities.

## Rotational Force

We investigated the motion of a particle in an optical trap but also with a rotational force.

The origin of this force could be some type of fluid flow or the incident light having some angular momentum component.

The form of the force we used is:

$$\mathbf{F} = - \begin{bmatrix} k_x & \gamma \omega \\ -\gamma \omega & k_y \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

$$= (-k_x x - \gamma \omega y) \hat{x} + (\gamma \omega x - k_y y) \hat{y}$$

Also we chose  $\omega$  such that  $\gamma \omega \sim k_x$  so all effects could still be observed.

Therefore as  $\gamma \sim 10^{-8}$  and  $k_x \sim 10^{-6}$  we chose  $\omega = 10^2$

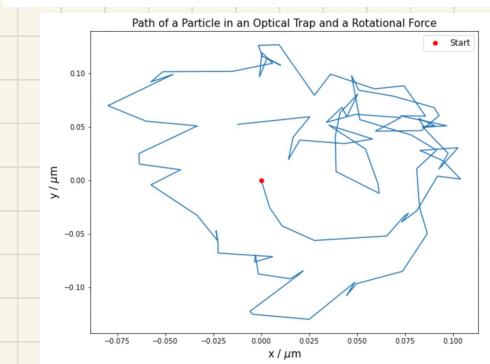
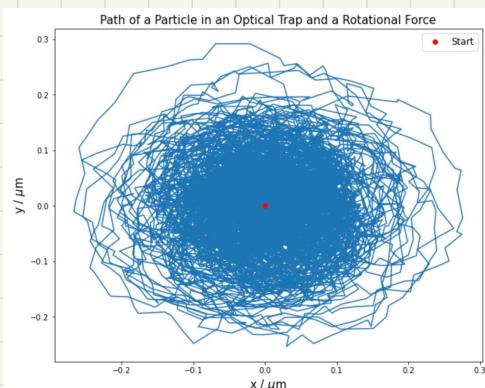
$$\gamma = 6\pi \eta n$$

$$10^{-6} \cdot 10^{-3}$$

$$10^{-9}$$

$$x_i = x_{i-1} + \frac{1}{\delta} [-k_x x_{i-1} - \gamma \omega y_{i-1}] \Delta t + \sqrt{2D \Delta t} \omega$$

$$y_i = y_{i-1} + \frac{1}{\delta} [-k_y y_{i-1} + \gamma \omega x_{i-1}] \Delta t + \sqrt{2D \Delta t} \omega$$



$$\gamma \omega x = [N]$$

$$N \text{sm}^{-2} [y] \text{m} = N$$

$$S \# [y] = 1 \quad y = \text{ans}$$

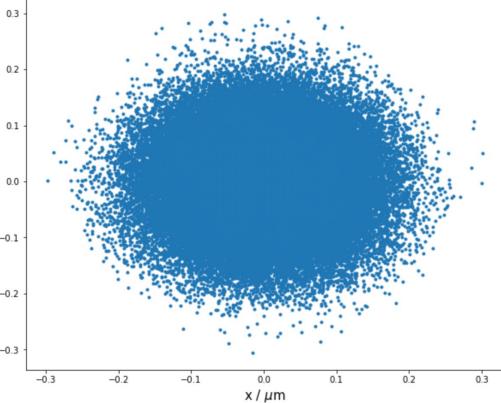
This is a plot of the first 100 steps.

It is difficult to see the effect of the rotational force with so many data points but it does show how the optical trap still works as a trap in the presence of a rotational force.

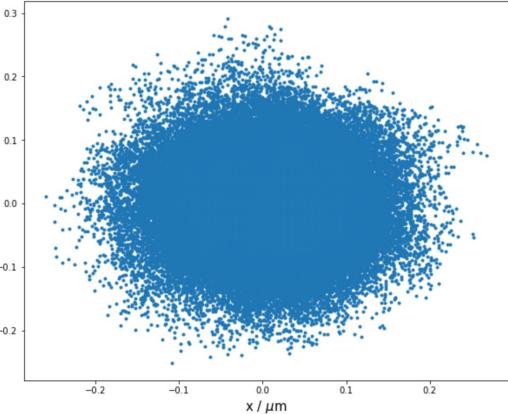
Plotting the same data but only the first 100 points shows the effect of the rotational force more clearly.

The particle gets pushed around anticlockwise but still takes random brownian motion steps. Also still confined by the trapping force.

Positions of a Particle in an Optical Trap and a Rotational Force

 $y / \mu\text{m}$ 

Positions of a Particle in an Optical Trap

 $y / \mu\text{m}$ 

The addition of a rotational force does not seem to have an effect on the effectiveness of the optical trap. The equilibrium position remains the same, the variation of the data is the same in  $x$  only.

## File Directories

All in folder: /Users/hamytakk/Dropbox/Uni/uni year 3/labs/lab twozzoo/

- Figures / tables used in lab interview ... / lab interview/
- Water Brownian motion ... / BM water/
- Equipartition method ... / EP method/
- Stokes method ... / Stokes/
- MSD method ... / Stiffness MSD/
- Varying Sugar concentration ... / Sugar/

All Simulation Code + Figures:

/Users/hamytakk/documents/labs/lab twozzoo/simulation/