



# 3D Particle Diffusions

Shiyu Liu

Advisors: Dr. Hoa Nguyen, Dr. Orrin Shindell

# Brownian Motion

Fluid ( $\eta$ )

Newton's Second Law:

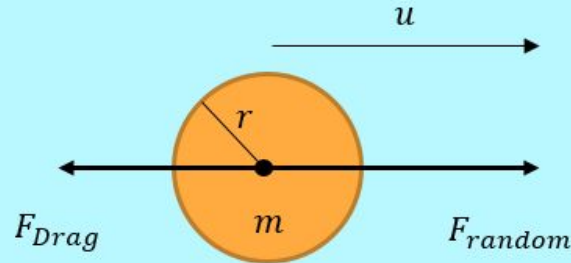
$$F_{net} = ma$$

For the Sphere:

$$m \frac{du}{dt} = -\gamma u + F_{random}$$

In Low Reynolds Number Flow:

$$\gamma = 6\pi\eta r$$

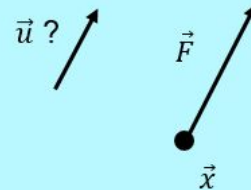


# Stoke Flows

- How to simulate the process of Brownian motion computationally

$$\begin{aligned}\mu \Delta \vec{u} &= \nabla p - \vec{F} \\ \nabla \cdot \vec{u} &= 0 \\ \Rightarrow \Delta p &= \nabla \cdot \vec{F}\end{aligned}$$

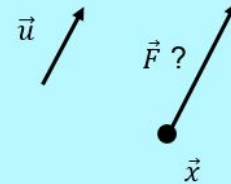
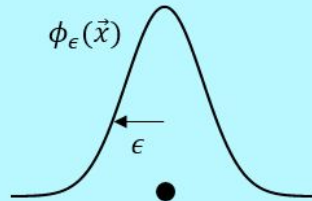
$\mu$  is the viscosity of the fluid,  $\vec{u}(\vec{x})$  is the velocity of the fluid,  $p(\vec{x})$  is the pressure,  $\vec{F}(\vec{x})$  is an external force, and  $(\vec{x})$  is a point in the fluid.  $\Delta \equiv \nabla \cdot \nabla$  is the Laplacian.



# Method of Regularized Stokeslets (MRS)

$$\vec{F}(\vec{x}) = \vec{f}_0 \phi_\epsilon(\vec{x} - \vec{x}_0)$$

$$\int \phi_\epsilon(\vec{x} - \vec{x}_0) d\vec{x} = 1$$



Cortez, R., Fauci, L., & Medovikov, A.S. (2005). *The method of regularized Stokeslets in three dimensions: Analysis, validation, and application to helical swimming*. Physics of Fluids, 17(3) 031504



Blob function  $\phi_{\epsilon}(\mathbf{x} - \mathbf{x}_0) = \frac{15\epsilon^4}{8\pi(r^2 + \epsilon^2)^{7/2}},$

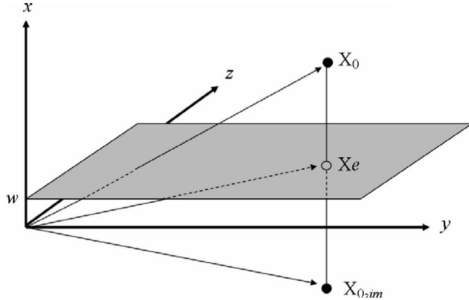
$$\mathbf{u}(\mathbf{x}) = \sum_{k=1}^N \frac{-\mathbf{f}_k}{4\pi\mu} \left[ \ln \left( \sqrt{r_k^2 + \epsilon^2} + \epsilon \right) - \frac{\epsilon \left( \sqrt{r_k^2 + \epsilon^2} + 2\epsilon \right)}{\left( \sqrt{r_k^2 + \epsilon^2} + \epsilon \right) \sqrt{r_k^2 + \epsilon^2}} \right] \\ + \frac{1}{4\pi\mu} [\mathbf{f}_k \cdot (\mathbf{x} - \mathbf{x}_k)] (\mathbf{x} - \mathbf{x}_k) \left[ \frac{\sqrt{r_k^2 + \epsilon^2} + 2\epsilon}{\left( \sqrt{r_k^2 + \epsilon^2} + \epsilon \right)^2 \sqrt{r_k^2 + \epsilon^2}} \right],$$

Due to the linear relationship between force and velocity, this formula can be written as

$$\mathbf{U} = \mathbf{MF}$$

# Method of Images for Regularized Stokeslets (MIRS)

$$\begin{aligned} \mathbf{U}(\mathbf{x}_e) = & \sum_{k=1}^M \left[ \mathbf{f}_k H_1(|\mathbf{x}_k^*|) + (\mathbf{f}_k \cdot \mathbf{x}_k^*) \mathbf{x}_k^* H_2(|\mathbf{x}_k^*|) \right] - \left[ \mathbf{f}_k H_1(|\mathbf{x}_k|) + (\mathbf{f}_k \cdot \mathbf{x}_k) \mathbf{x}_k H_2(|\mathbf{x}_k|) \right] - h_k^2 [\mathbf{g}_k D_1(|\mathbf{x}_k|) \\ & + (\mathbf{g}_k \cdot \mathbf{x}_k) \mathbf{x}_k D_2(|\mathbf{x}_k|)] - 2h_k \left[ \frac{H'_1(|\mathbf{x}_k|)}{|\mathbf{x}_k|} + H_2(|\mathbf{x}_k|) \right] (\mathbf{L}_k \times \mathbf{x}_k) + 2h_k \left[ (\mathbf{g}_k \cdot \mathbf{e}_1) \mathbf{x}_k H_2(|\mathbf{x}_k|) \right. \\ & \left. + (\mathbf{x}_k \cdot \mathbf{e}_1) \mathbf{g}_k H_2(|\mathbf{x}_k|) + (\mathbf{g}_k \cdot \mathbf{x}_k) \mathbf{e}_1 \frac{H'_1(|\mathbf{x}_k|)}{|\mathbf{x}_k|} + (\mathbf{x}_k \cdot \mathbf{e}_1) (\mathbf{g}_k \cdot \mathbf{x}_k) \mathbf{x}_k \frac{H'_2(|\mathbf{x}_k|)}{|\mathbf{x}_k|} \right], \end{aligned}$$



Schematic of the fluid domain with the wall at  $x = w$ , the location of the Stokeslet at  $\mathbf{x}_0$  and the image point.

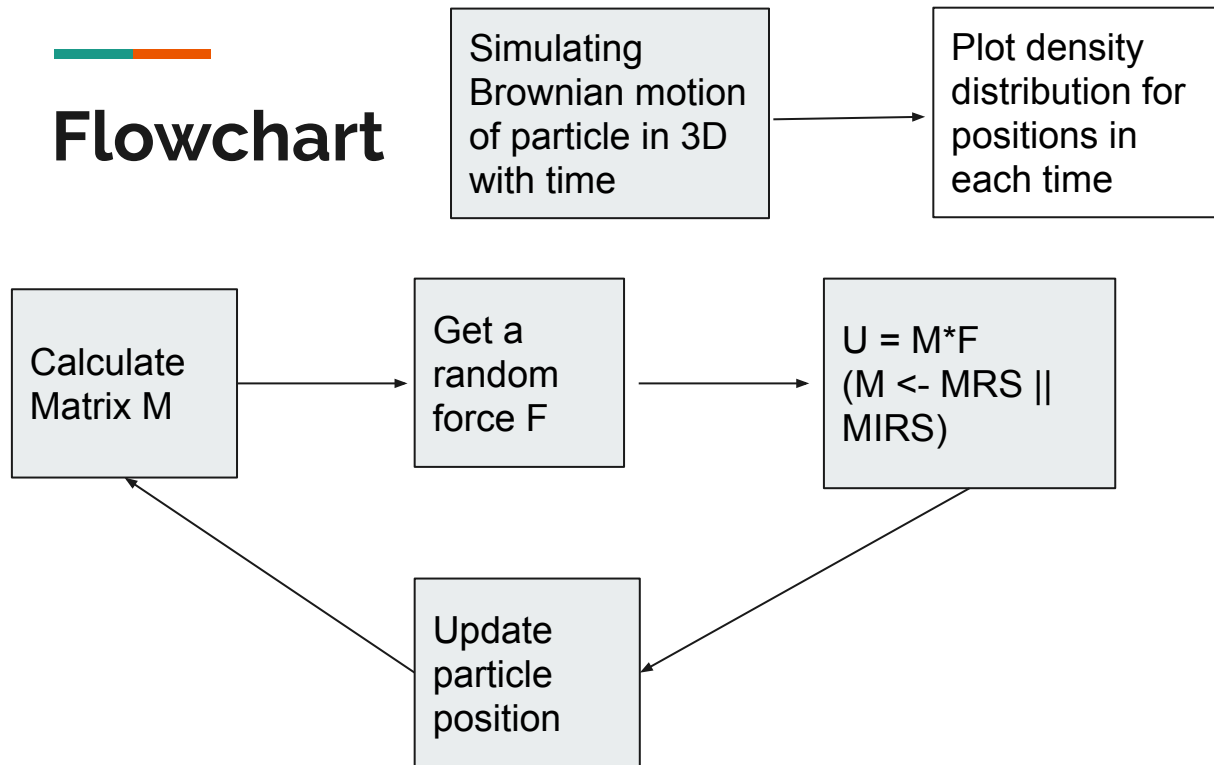
Due to the linear relationship between force and velocity, this formula can be written as

$$\mathbf{U} = \mathbf{M}\mathbf{F}$$

Ainley, J., Durkin, S., Embid, R., Boindala, P., & Cortez, R. (2008). *The method of images for regularized Stokeslets*. Journal of Computational Physics, 227(9), 4600–4616.



# Flowchart



***N particles/simulation***  
vs.  
***1 particle/simulation***  
***for N simulations***

***MRS vs. MIRS***

## 3-D Task 1: Find Drag Radius as a Function of $\epsilon$

- Impose a constant velocity  $\vec{u}$  to a blob of size  $\epsilon$ , find the drag force  $\vec{F}$ , and use the Stokes Drag Formula  $F = (6\pi\eta r)u$  ...

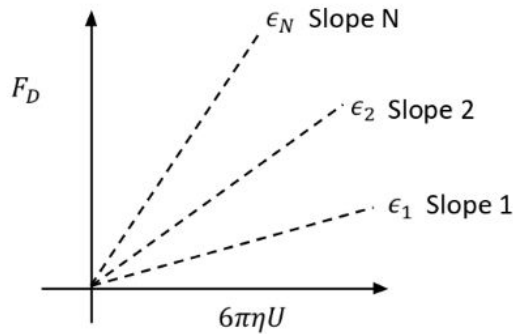


Figure 1.  $F_D$  versus  $6\pi\eta U$  for various values of  $\epsilon$ .

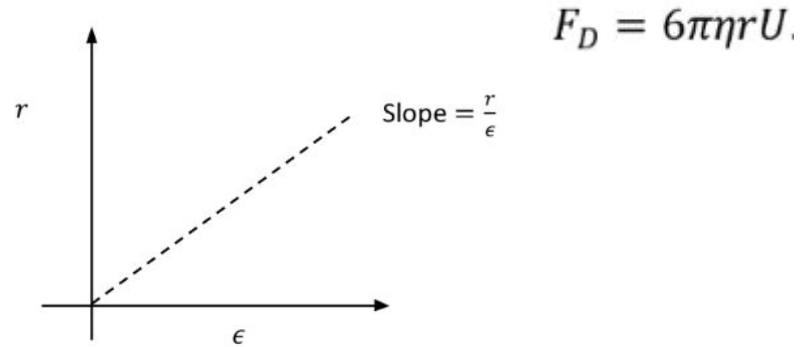
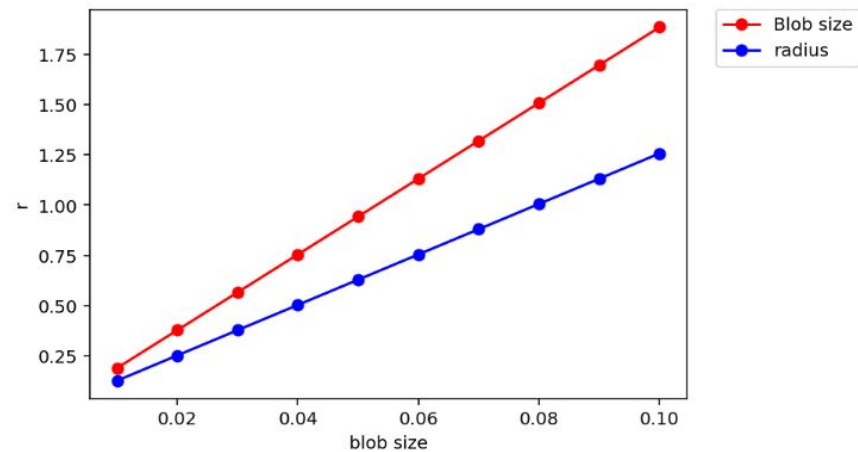
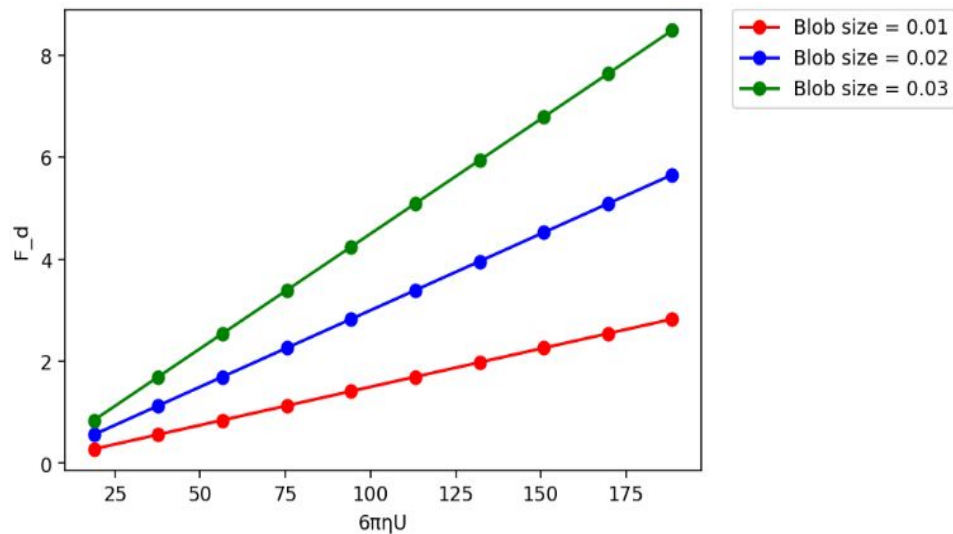


Figure 2.  $\frac{F_D}{6\pi\eta U} = r$  versus  $\epsilon$ . The slope gives a scaling from  $\epsilon$  to  $r$ .



# Result

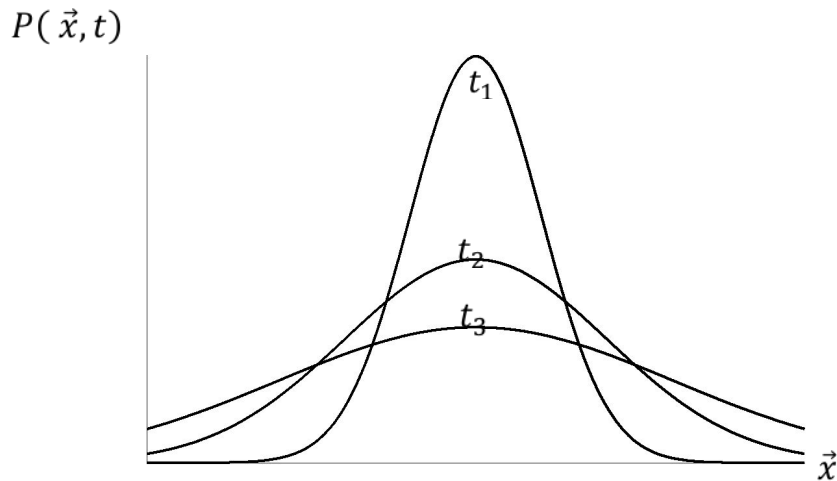
Blob size  $\mathcal{E} = 1.5r$  where  $r$  is the radius of a particle.



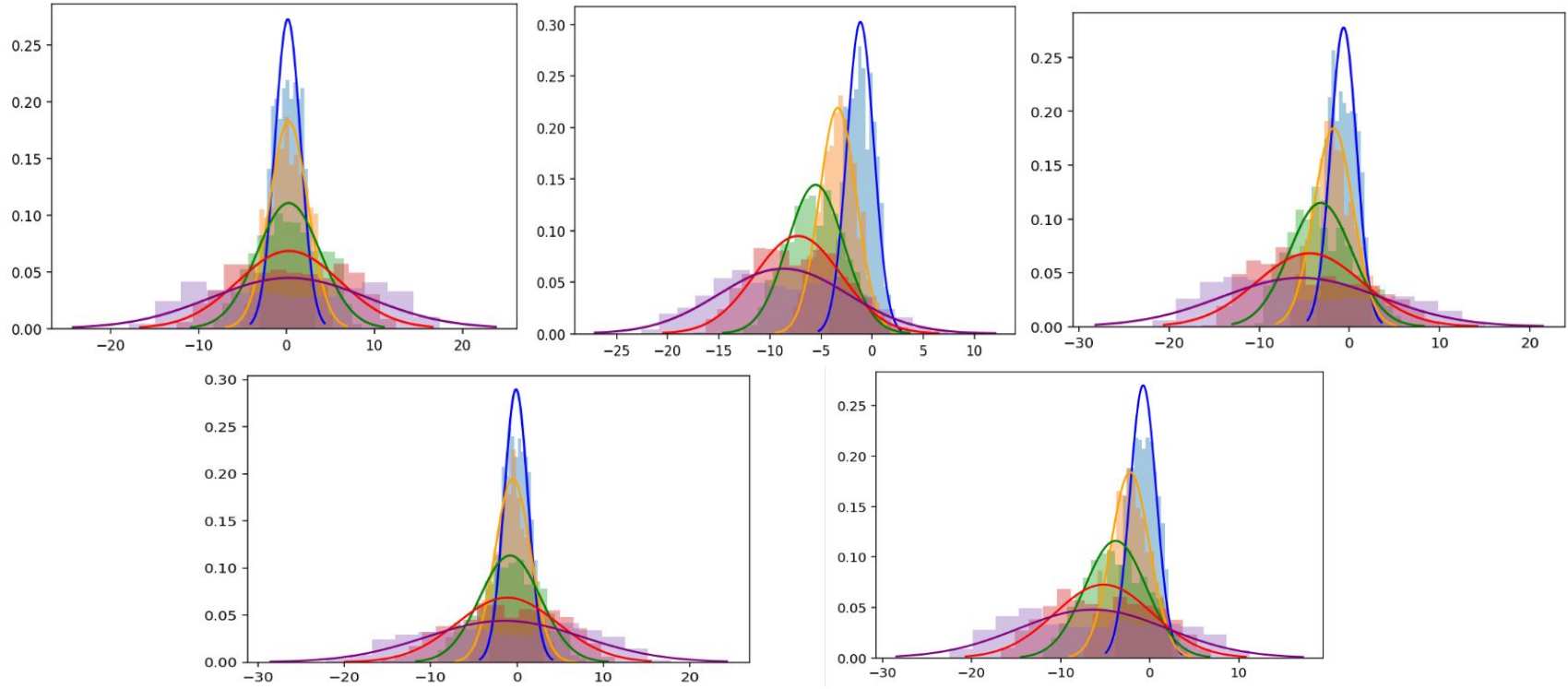
## 2. Simulate Brownian Motion

- Add in a random force at each time step, calculate the probability density of the particle being found at position  $\vec{x}$  at time  $t$ .

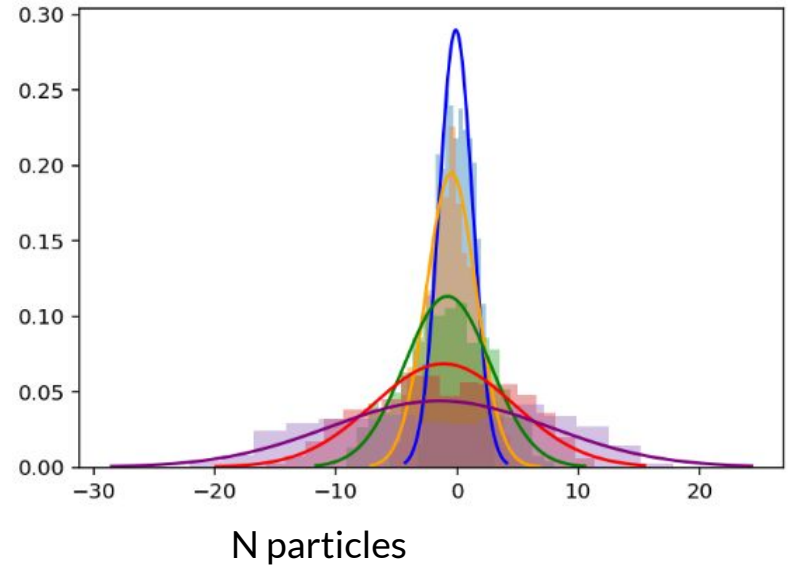
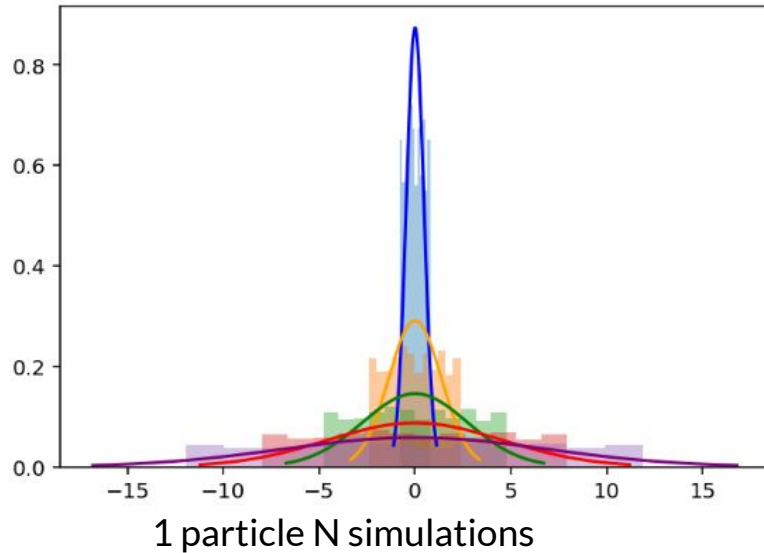
$$P(\vec{x}, t) = \frac{1}{\sqrt{(4\pi Dt)^3}} e^{-\frac{\vec{x}^2}{4Dt}}$$



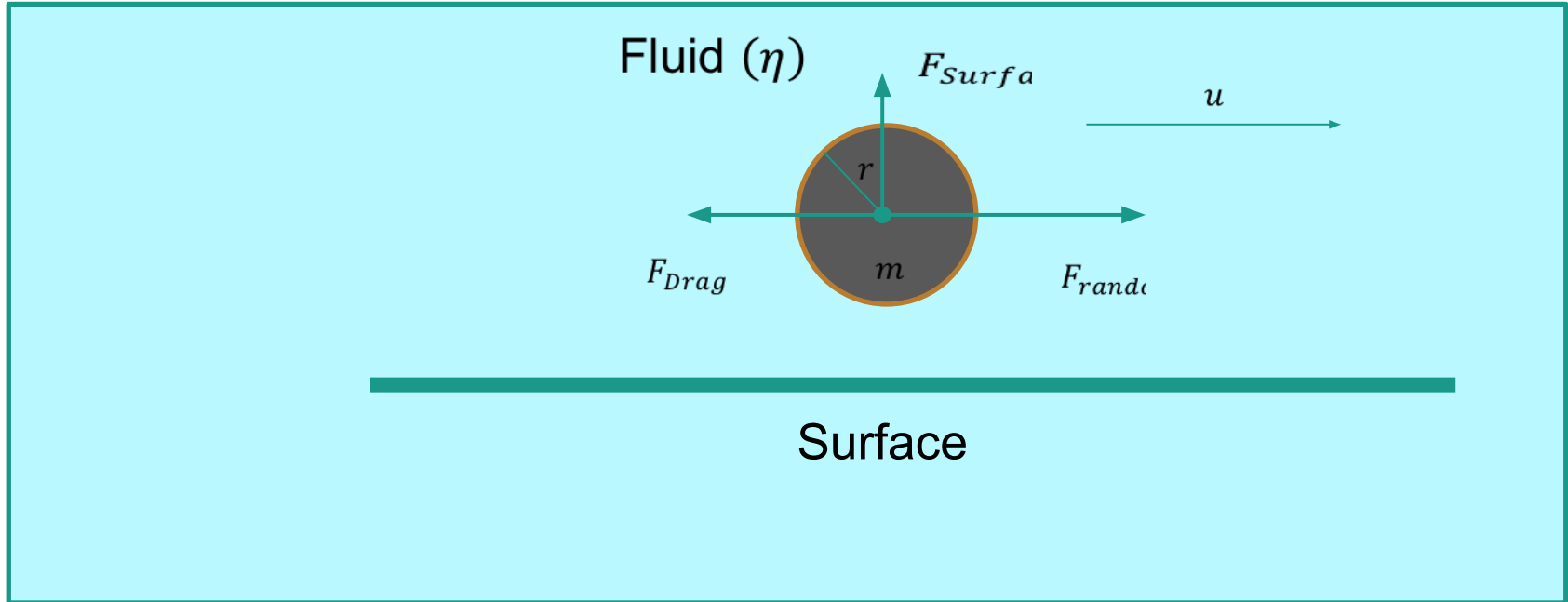
# Results for N particles/simulation using MRS (no surface)



# Results for 1 particle/simulation with N simulations using MRS (no surface)

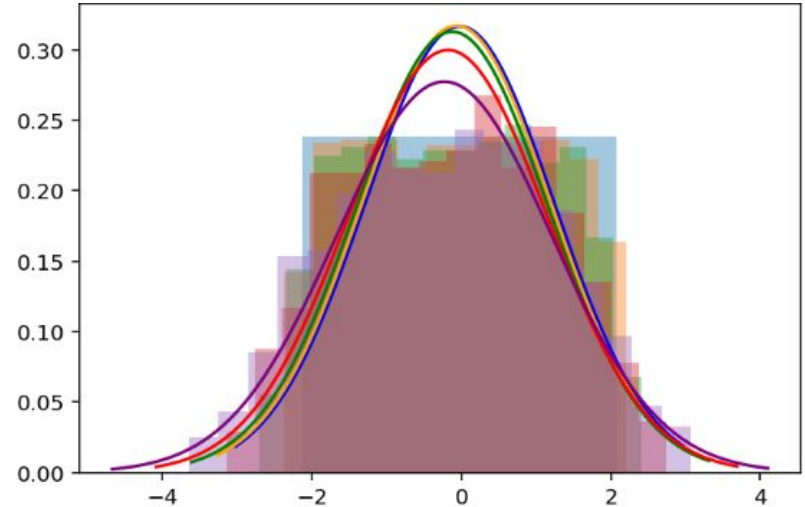
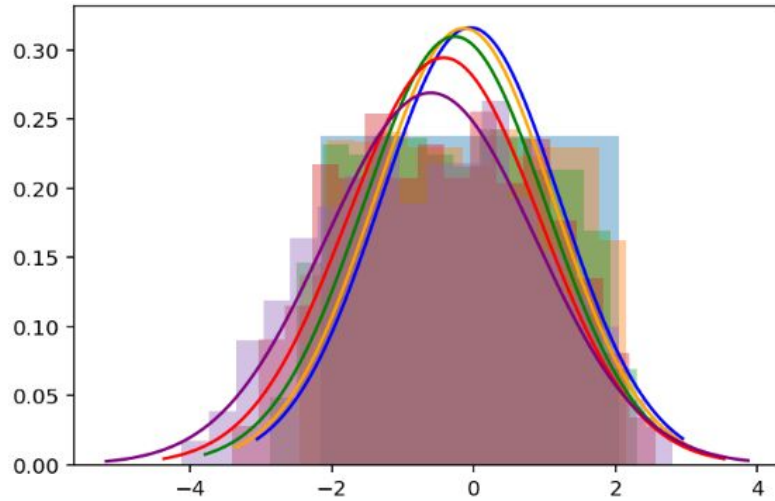


### 3-D Task 3: Impose Surface Hydrodynamic Interactions



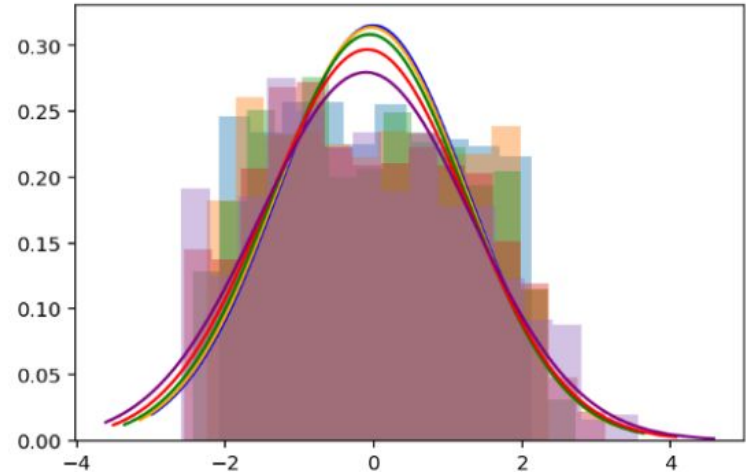
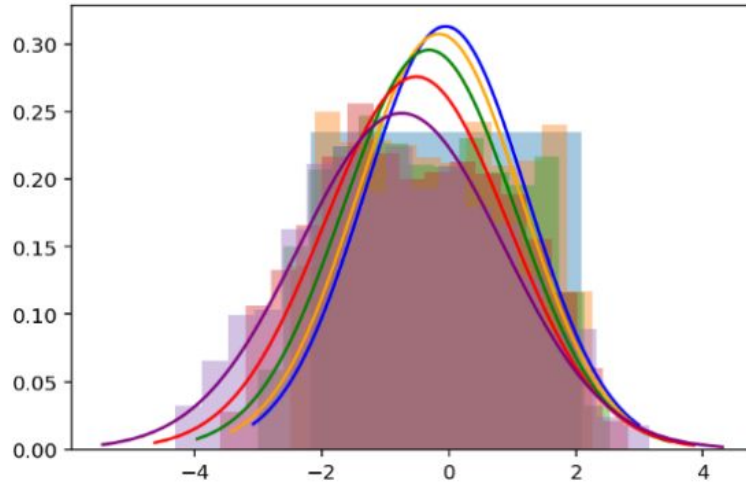
# Results for N particles/simulation using MIRS

(density distribution in the y; surface at  $-2.5 x$ )

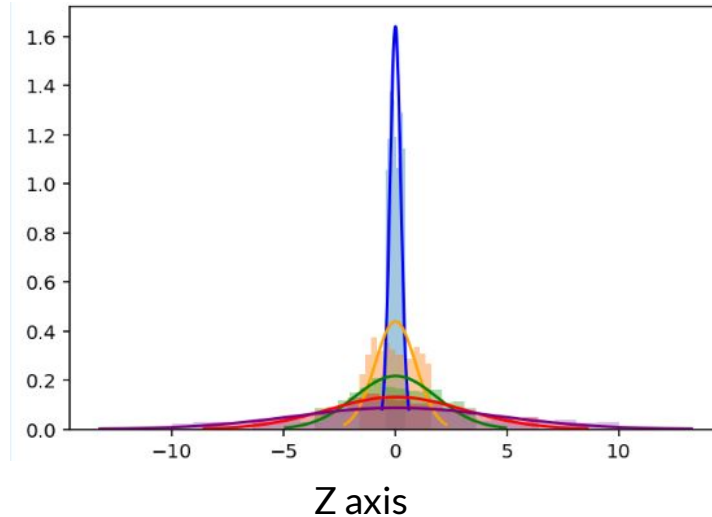
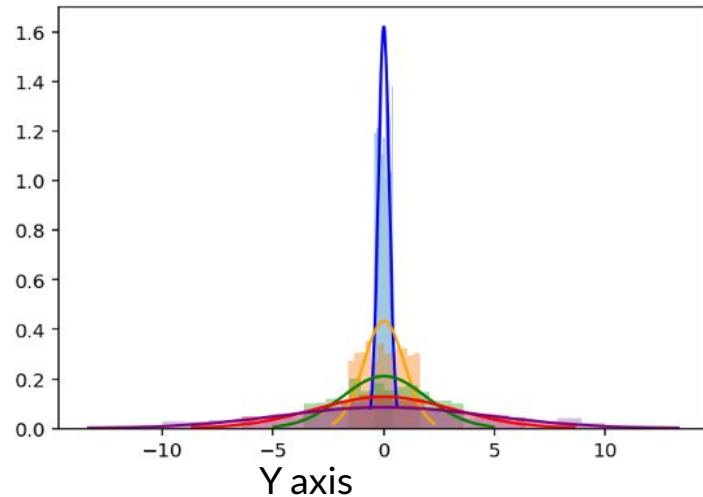


## Results for N particles/simulation using MIRS

(density distribution in the x direction; surface at  $x = -2.5$ )

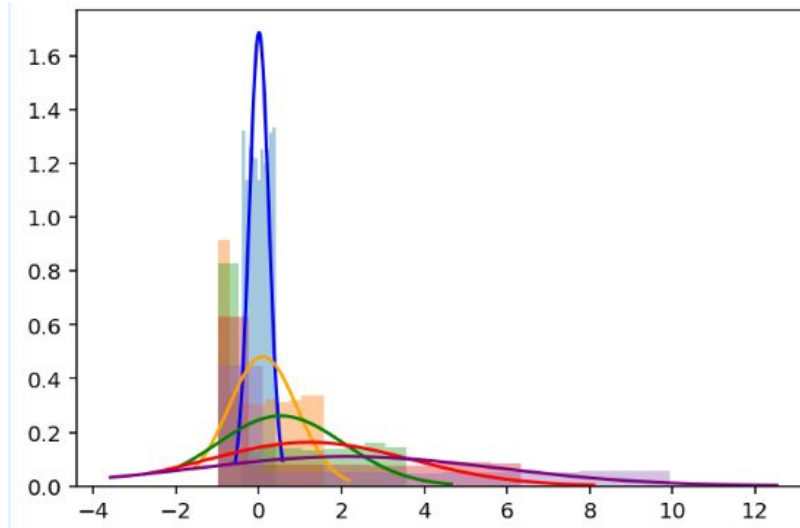


# Results for 1 particle/simulation with N simulations using MIRS (density distribution in the y or z direction; surface at $x = -1$ )



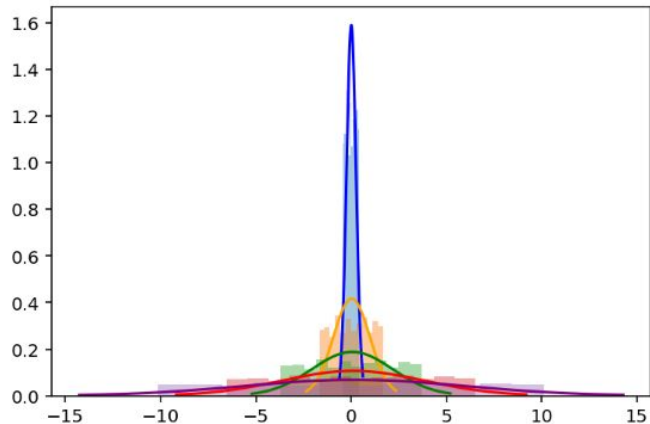


# Results for 1 particle/simulation with N simulations using MIRS (density distribution in the x direction; surface at x = -1)

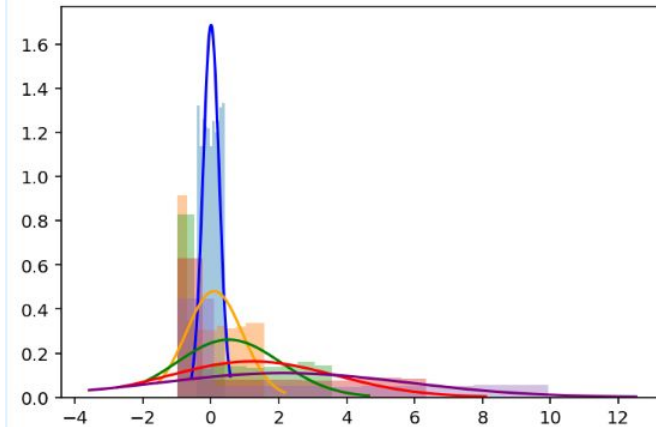


# Conclusion 1: Surface Effect (No Particle-Particle Interactions)

1 particle/simulation for N simulations (density distribution in the x direction)



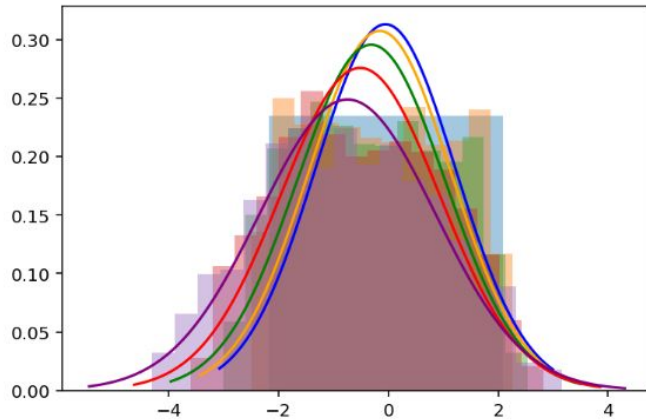
MRS (infinite 3D space)



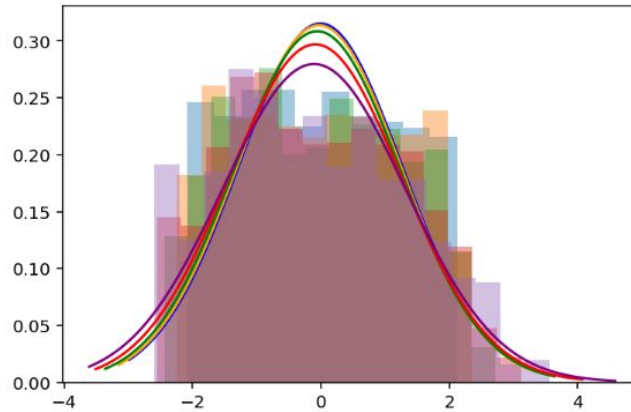
MRS (surface at  $x = -1$ )

## Conclusion 2: Surface Effect *and* Particle-Particle Interactions

N particles/simulation (density distribution in the x direction)



MRS (infinite 3D space)



MIRS (surface at  $x = -2.5$ )



# Lessons & Future directions

- Time management
- Communication with advisors regularly
- Similarity to machine learning

## Future directions

- Realistic random force
- Further applications



# Acknowledgement

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- Dr. Shindell
- Dr. Nguyen
- Dr. Lewis for coming today!