#### How swimming microorganisms displace fluid particles

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[play movie](http://www.math.wisc.edu/~jeanluc/movies/Guasto2010_start.mp4)

[Guasto, J. S., Johnson, K. A., & Gollub, J. P. (2010). Phys. Rev. Lett. 105, 168102]

# Probability density of displacements

Non-Gaussian PDF with 'exponential' tails:



[Leptos, K. C., Guasto, J. S., Gollub, J. P., Pesci, A. I., & Goldstein, R. E. (2009). Phys. Rev. Lett. 103, 198103]



Leptos et al. (2009) get a reasonable fit of their PDF with the form

$$
\mathbb{P}\{X_t \in [x, x + \mathrm{d}x]\} = \frac{1 - f}{\sqrt{2\pi \delta_g^2}} e^{-x^2/2\delta_g^2} + \frac{f}{2\delta_e} e^{-|x|/\delta_e}.
$$

They observe the scalings  $\delta_{\rm g}\approx A_{\rm g}t^{1/2}$  and  $\delta_{\rm e}\approx A_{\rm e}t^{1/2}$ , where  $A_{\rm g}$  and  $A_{\rm e}$ depend on the volume fraction  $\phi$ .

They call this a diffusive scaling, since  $\lambda_t/t^{1/2}$  is a scaling variable. Their point is that this is strange, since the distribution is not Gaussian.

Commonly observed in diffusive processes that are a combination of trapped and hopping dynamics (Wang *et al.*, 2012).

### Modeling: the interaction sphere





Model for effective diffusivity:

[Thiffeault, J.-L. & Childress, S. (2010). Phys. Lett. A, 374, 3487–3490]

[Lin, Z., Thiffeault, J.-L., & Childress, S. (2011). J. Fluid Mech. 669, 167–177]

Expected number of 'dings' (close interactions) after time t:

$$
\langle M_t \rangle = n \{ V_{\text{swept}}(R, \lambda) (t/\tau) + V_{\text{sph}}(R) \}
$$

*n* is the number density of swimmers,  $V_{\text{swent}}$  is the volume swept by the sphere of radius R moving a distance  $\lambda$ , and  $\tau$  is the time between turns.

#### Parameters in the Leptos et al. experiment

- Velocity  $U \sim 100 \,\mu{\rm m/s}$ ;
- Volume fraction is less than 2.2%;
- Organisms of radius  $5 \mu m$ ;
- Number density  $n \lesssim 4.2 \times 10^{-5} \ \mu \mathrm{m}^{-3}$ .
- Maximum observation time in PDFs is  $t \sim 0.3$  s;
- A typical swimmer moves by a distance  $Ut \sim 30 \,\mu \mathrm{m}$ .

Combining this, we find the expected number of 'dings' after time  $t$  in the Leptos et al. experiment:

$$
\langle M_t \rangle \lesssim 0.6
$$

for the longest observation time, and interaction sphere  $R = 10 \,\mu \mathrm{m}$ .

Conclude: a typical fluid particle is only strongly affected by about one swimmer during the experiment.

The only displacements that a particle feels 'often' are the very small ones due to all the distant swimmers.

We thus expect the displacement PDF to have a central Gaussian core (since the central limit theorem will apply for the small displacements), but strongly non-Gaussian tails.



- $\bullet$   $X_t$  is the displacement of a particle after a time  $t$ ;
- $X_m$  is the displacement of a particle after m encounters;
- But the number of encounters is a random variable  $M_t$ .
- How do we relate the two?

$$
\mathbb{P}\{X_t \in [x, x + dx]\} = \sum_{m=0}^{\infty} \mathbb{P}\{X_t \in [x, x + dx], M_t = m\}
$$
  
= 
$$
\sum_{m=0}^{\infty} \mathbb{P}\{X_t \in [x, x + dx] | M_t = m\} \mathbb{P}\{M_t = m\}
$$
  
= 
$$
\sum_{m=0}^{\infty} \mathbb{P}\{X_m \in [x, x + dx]\} \mathbb{P}\{M_t = m\}
$$

When the volume is large, the number of interactions obeys a Poisson distribution:

$$
\mathbb{P}\{M_t = m\} \simeq \frac{1}{m!} \langle M_t \rangle^m e^{-\langle M_t \rangle}
$$

We define the probability densities:

$$
\rho_{X_m}(x) dx := \mathbb{P}\{X_m \in [x, x + dx]\}
$$

$$
\rho_{X_t}(x) dx := \mathbb{P}\{X_t \in [x, x + dx]\}
$$

From previous slide:

$$
\rho_{X_t}(x) = \sum_{m=0}^{\infty} \rho_{X_m}(x) \mathbb{P}\{M_t = m\}
$$



Normally we would now go to the large  $m$  limit and use large-deviation theory. But this doesn't hold here. Instead, keep only  $m \leq 1$ ,

$$
\rho_{X_t}(x) = \sum_{m=0}^{\infty} \rho_{X_m}(x) \mathbb{P}\{M_t = m\}
$$
  
 
$$
\simeq \mathbb{P}\{M_t = 0\} \rho_{X_0}(x) + \mathbb{P}\{M_t = 1\} \rho_{X_1}(x) + \dots
$$

i.e., most fluid particles feel only a few close encounters with swimmers.

 $\rho_{\mathcal{X}_0}(\mathsf{x})$  is due to thermal noise (or the combined effect of distant swimmers), so is Gaussian.

 $\rho_{\boldsymbol{\mathsf{X}}_1}(\text{x})$  is the displacement probability after one close interaction with a swimmer, which has strongly non-Gaussian tails.



#### Geometry of an encounter





# The single-encounter probability  $\rho_{X_1}(x)$

We can show that (Thiffeault, 2014)

$$
\rho_{X_1}(x) = \frac{1}{2} \int_{\Omega_{ab}} \frac{\rho_{AB}(a, b)}{\Delta_{\lambda}(a, b)} \chi_{\{\Delta_{\lambda} > |x|\}}(a, b) \, da \, db,
$$

where

- a and  $b$  are the impact parameters that describe the geometry of an encounter;
- $\Delta_{\lambda}$  is the drift function;
- $\chi$  is an indicator function (i.e., 0 or 1);
- $\rho_{AB}(a, b) = 2\pi a/V_{\text{swept}}(R, \lambda)$  is the probability density of the random impact parameters  $A$  and  $B$ .

The drift function is computed (laboriously) by integrating over fluid trajectories.

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[Thiffeault, J.-L. (2014). arXiv:1408.4781]
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What about the density function for two encounters,  $\rho_{X_2}(x)$ ?

Since  $X_2$  is the sum of two i.i.d. random variables  $X_1$ , its PDF is just the convolution of  $\rho_{X_1}(x)$  with itself:

$$
\rho_{X_2}(x) = \int_{-\infty}^{\infty} \rho_{X_1}(x - y) \, \rho_{X_1}(y) \, \mathrm{d}y =: (\rho_{X_1} * \rho_{X_1})(x).
$$

For *m* steps we have  $\rho_{X_m}(x) = (\rho_{X_1} * \cdots * \rho_{X_1})(x)$ .

[The central limit theorem / large deviation theory are estimates of this convolution for large m.]

### A model swimmer



This is as far as we can go without introducing a model swimmer.

We take a squirmer, with axisymmetric streamfunction:

$$
\Psi_{\text{sf}}(\rho, z) = \frac{1}{2}\rho^2 U \left\{ -1 + \frac{\ell^3}{(\rho^2 + z^2)^{3/2}} + \frac{3}{2} \frac{\beta \ell^2 z}{(\rho^2 + z^2)^{3/2}} \left( \frac{\ell^2}{\rho^2 + z^2} - 1 \right) \right\}
$$

[See for example Lighthill (1952); Blake (1971); Ishikawa et al. (2006); Ishikawa & Pedley (2007b); Drescher et al. (2009)]

We use the stresslet strength  $\beta = 0.6$ , which is close to a treadmiller:



# $\rho_{X_m}\!(x)$  for the squirmer





#### Comparing to Leptos et al.





The only fitted parameter is the stresslet strength  $\beta = 0.6$ .

# Comparing to Eckhardt & Zammert

Eckhardt & Zammert (2012) have a beautiful fit to the data based on a phenomenological continuous-time random walk model (dashed):



Our models disagree in the tails, but there is no data there.

What about the 'diffusive scaling' mentioned at the start?







It's present in our model as well:



(Earlier times are a bit worse.)



It persists (except for cut-off) further in the tails:



Note that the times are still short enough that the organisms don't have time to turn.



#### Appears to hold for a single encounter, for  $\rho_{X_1}(x)$ :



This means the scaling is not really statistical in nature: it's a property of the drift function  $\Delta_{\lambda}$  itself for this type of swimmer.

If we go further in time and allow the organisms to reorient, the scaling seems to disappear completely:





- Times in Leptos et al. (2009) are so short that the tails are not determined by asymptotic laws, such as the central limit theorem or large-deviation theory.
- Retaining only 0 and 1 close interactions gives a linear combination of a Gaussian and a distribution with non-Gaussian tails, as observed by Leptos et al. (2009).
- The Gaussian core arises because of the net effect of the distant swimmers, far from the test particle.
- Preprint: <http://arxiv.org/abs/1408.4781>.

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