## Magnetic Navigation of Artificial Bacteria Flagella in Blood and Water

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# Artificial Microswimmers



Pieters, Roel S., et al. "Non-contact manipulation for automated protein crystal harvesting using a rolling microrobot." IFAC Proceedings Volumes 47.3 (2014): 7480-7485.

Medina-Sánchez, Mariana, et al. "Cellular cargo delivery: Toward assisted fertilization by sperm-carrying micromotors." Nano letters 16.1 (2016): 555-561.

# Outline

1. Independent Control and Path Planning of artificial bacterial flagella

2. Artificial bacterial flagella swimming in blood

# Artificial Bacterial Flagella (ABF)

## Regime 1

## Re $\approx 10^{-4} \ll 1 \Rightarrow$ Stokes flow. $\begin{bmatrix} \mathbf{V}^B \\ \mathbf{\Omega}^B \end{bmatrix} = \begin{bmatrix} \mathscr{A} & \mathscr{B} \\ \mathscr{B} & \mathscr{C} \end{bmatrix} \begin{bmatrix} \mathbf{F}^B \\ \mathbf{T}^B \end{bmatrix}$ Magnetic torque: $\mathbf{T} = \mathbf{m} \times \mathbf{B}$

## ODE model

$$\dot{q} = \frac{1}{2} q \otimes \hat{\Omega}, \qquad \mathbf{V}^{B} = \mathscr{B} \mathbf{T}^{B},$$
$$\dot{\mathbf{x}} = \mathbf{V}, \qquad \mathbf{\Omega}^{B} = \mathscr{C} \mathbf{T}^{B}.$$





# Different geometries allow independent control





The geometry governs the response to the rotation frequency of the magnetic field





# Swimming different distances along a direction

## **Constraint: All swimmers are subjected to the same magnetic field**



(signed) distance made by sw

where  $s_i = -1$  (clockwise) o

 $\Rightarrow \mathbf{b} = U^{-1}\mathbf{d}$ 

Alternate the magnetic field rotation frequency:

vimmer 
$$i: d_i = \sum_j U_{ij} t_j s_j = U_{ij} b_j$$
  
r  $s_j = +1$  (anti-clockwise)



# Gathering swimmers to a target

 $\mathbf{k}_1$ 

 $\mathbf{k}_{2}$ 

In 3 dimensions: 3 steps to reach the target

- Gather on a plane 1.
- 2. Gather on a line
- 3. Gather at the target



## Independent control in free space



## Reinforcement Learning





## Independent control in free space: RL





# Independent control with a background flow

Taylor-Green stationary flow

 $\mathbf{u}_{\infty}(\mathbf{r}) =$ 

 $\begin{bmatrix} A \cos ax \sin by \sin cz \\ B \sin ax \cos by \sin cz \end{bmatrix}$ 

 $C\sin ax\sin by\cos cz$ 







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## Blood Model

**Bending Energy** 

$$E_b = 2\kappa_b \oint \left(H - H_0\right)^2 dA$$

Jülicher, F. 1996. Journal de Physique II, 6(12), 1797–1824.

### **Dissipation forces**

$$\mathbf{f}_{i}^{visc} = -\sum_{j} \gamma \left( \mathbf{v}_{ij} \cdot \mathbf{e}_{ij} \right) \mathbf{e}_{ij}$$

Fedosov, et al. 2010. Biophysical Journal, 98(10), 2215–2225.

### Area and Volume penalization

$$E_A = k_A \frac{(A - A_0)^2}{A_0}$$
,  $E_V = k_V \frac{(V - V_0)^2}{V_0}$ 

Fedosov, et al. 2010. Biophysical Journal, 98(10), 2215–2225.

### Shear Energy

AVAVAVA.

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$$E_s = \frac{K_\alpha}{2} \oint \left(\alpha^2 + a_3 \alpha^3 + a_4 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha \beta + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha + b_2 \alpha^4\right) dA_0 + \mu \oint \left(\beta + b_1 \alpha + b_2 \alpha^4\right) dA_0$$

with respect to stress-free shape:

Lim et al. 2008. Soft Matter, 4.



## Solvent Model



Bounce Back on the membrane

$$\mathbf{v}(t + dt) = 2\mathbf{v}_{RBC}(t_{collision}) - \mathbf{v}(t)$$

**Dissipative Particle Dynamics interactions** 

$$\begin{aligned} \mathbf{f}_{ij}^{C} &= aw(r_{ij})\mathbf{e}_{ij}, & \text{hydrostatic pressure} \\ \mathbf{f}_{ij}^{D} &= -\gamma w_{D}(r_{ij}) \left(\mathbf{e}_{ij} \cdot \mathbf{v}_{ij}\right) \mathbf{e}_{ij}, & \text{viscosity} \\ \mathbf{f}_{ij}^{R} &= \sigma \xi_{ij} w_{R}(r_{ij}) \mathbf{e}_{ij} & \text{fluctuations} \end{aligned}$$



# ABFs swimming in blood

### Ht = 10%



## Ht = 20%



## ABFs swim faster at higher hematocrit





# ABF catching a circulating cancerous cell

## Uncontrolled

Ht = 20%



## Controlled

## Ht = 20%



Controlled to stay near the center of the pipe









 $\alpha$ 

# Summary

- Reinforcement learning is a good tool for independent control of multiple swimmers under a uniform magnetic field
- Artificial bacterial flagella can navigate in blood efficiently if the magnetic torque is high enough
- Opens the road to optimize the design of single and swarms of swimmers





