

A microscopic image of cells, likely from a tissue sample. The nuclei are stained blue, and the cytoplasm and extracellular matrix show green and red fluorescence. The image is overlaid with a grid pattern.

Imperial College
London

Collective dynamics and self organization in biological systems

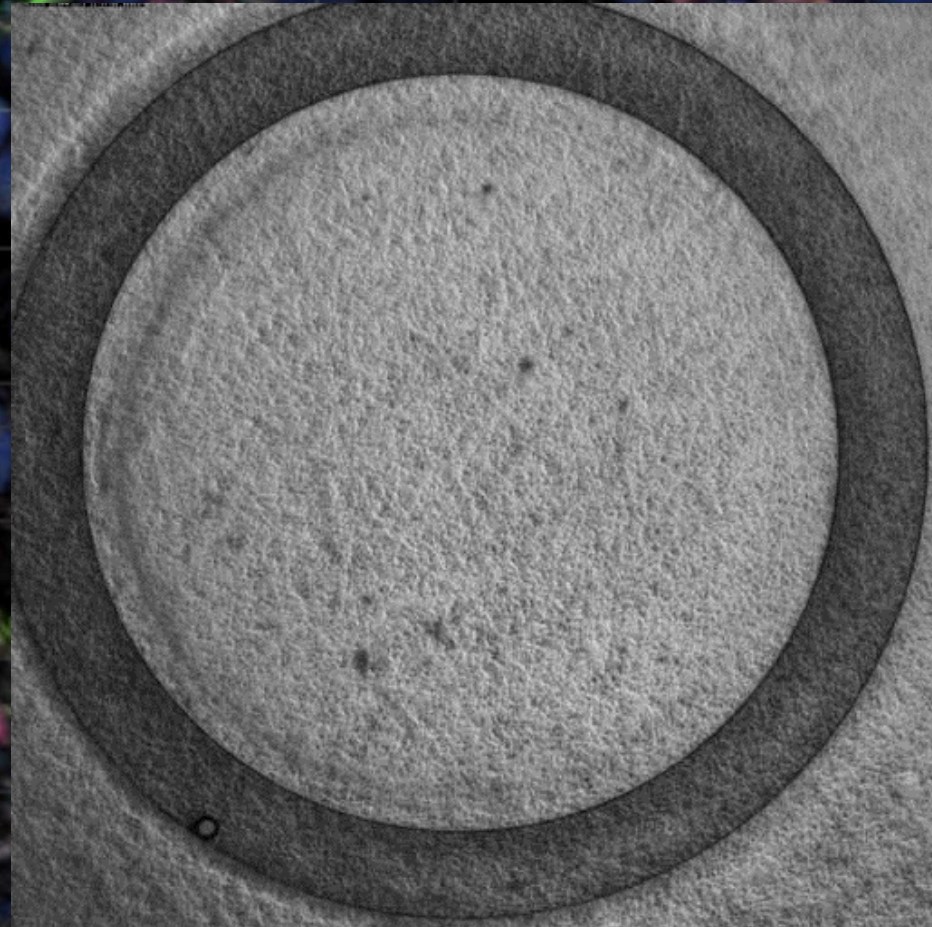
Lecture 1: Challenges and some examples

Pierre Degond
Imperial College London

What does “collective dynamics” mean ?

Coordination

Ex. all particles move spontaneously in the same direction



F. Plouraboué



X. Druart



H. Yu

Sperm confined in an annular chamber.
Creppy, Plouraboué, Praud, Druart, Cazin, Yu,
PD, J. Roy Soc. Interface 2016

What does “collective dynamics” mean ?

Coordination

Self-organization

Ex. spontaneous lane
formation



M. Moussaïd



J. Pettré



C. Appert-Rolland



G. Theraulaz

Pedestrians walking in an annular corridor.

Moussaïd, Guillot, Moreau, Fehrenbach, Chabiron, Lemerrier, Pettré,
Appert-Rolland, PD, Theraulaz, PLoS CB, 8 (2012), e1002442

What does “collective dynamics” mean ?

Coordination

Self-organization

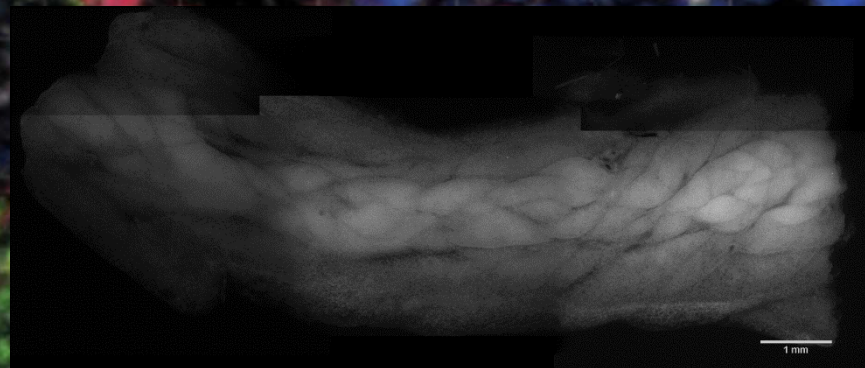
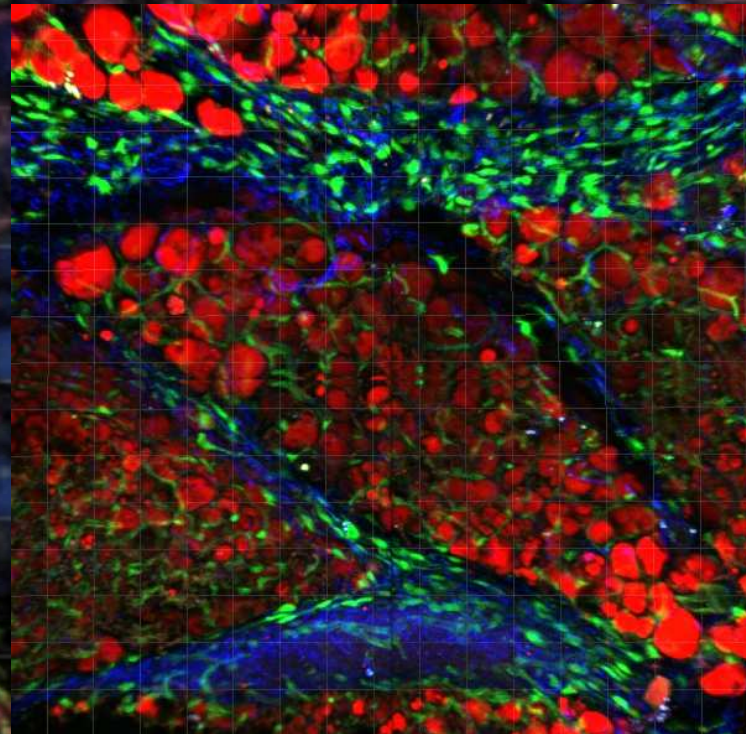
Pattern formation

Ex. Network emergence



L. Casteilla

Network of collagen
fibers in adipose tissue.
Courtesy of L. Casteilla
& A. Lorsignol



What does “collective dynamics” mean ?

Coordination

Self-organization

Pattern formation



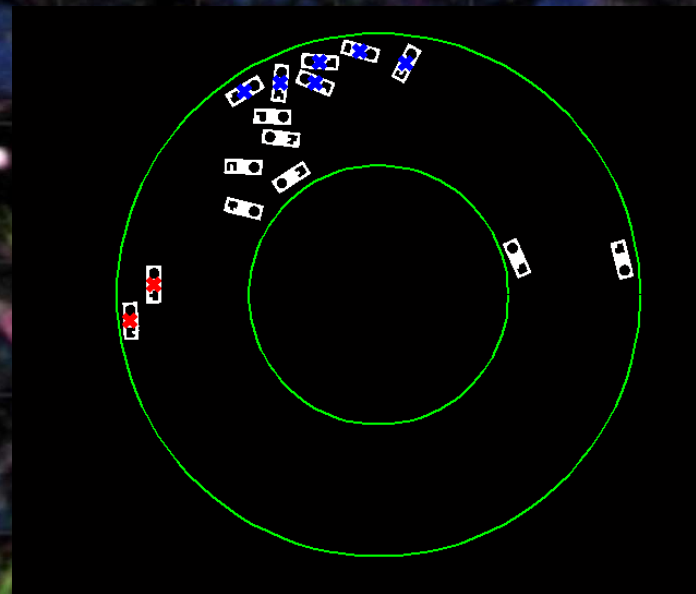
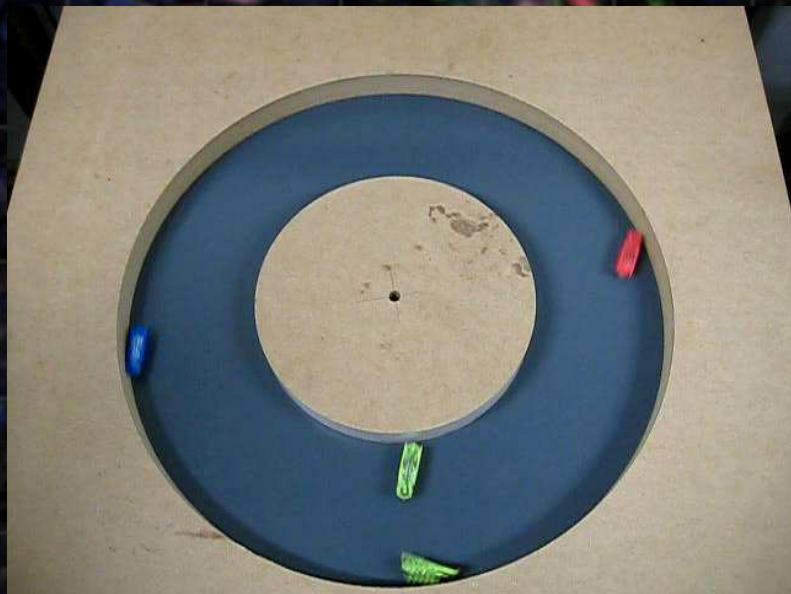
Emergent behavior

Not encoded in the
individual particle
interactions:

“complex systems”

What does “collective dynamics” mean ?

Systems showing emergent behavior **do not exclusively come** from biology or social sciences



With
E. Climent,
N. Mac,
F. Plouraboué,
O. Praud,

E. Climent

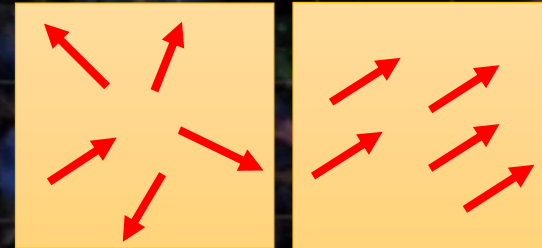


F. Plouraboué

Why is studying emergence difficult?

Emergence = Transition:
disorder \rightarrow self-organization

Ex. random state vs
aligned state



A. Frouvelle



Depends on noise

i.e. how often particles
change orientation
randomly

Ex. Vicsek model

self-propulsion +
alignment + noise

Vicsek, Czirók, Ben-
Jacob, Cohen, Shochet,
PRL 75 (1995) 1226

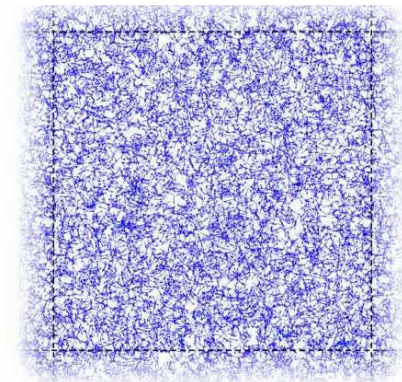
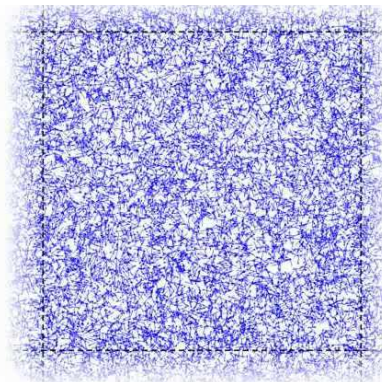


T. Vicsek

Alignment interaction + noise
Simulation by A. Frouvelle

Larger noise

Smaller noise

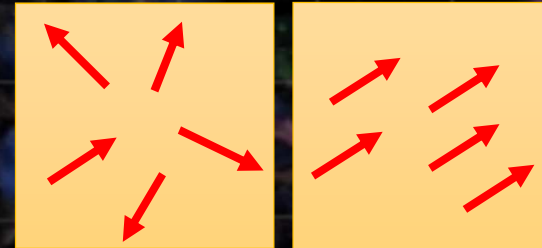


t = 00,00

Why is studying emergence difficult?

Emergence = Transition:
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Ex. random state vs
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Depends on noise

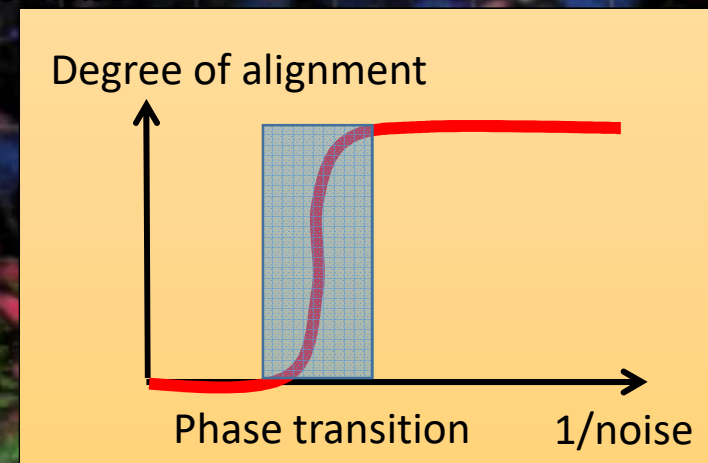
In an abrupt way

All variation in narrow parameter
range

Phase transition (or bifurcation)

Non-smooth behavior !

Alignment interaction + noise
Simulation by A. Frouvelle
Larger noise Smaller noise



Different types of phase transitions

Symmetry breaking

Ex. isotropic to polarized

Groups of *Khulia mugil*
in tank. Courtesy of G.
Theraulaz et al



G. Theraulaz



Different types of phase transitions

Symmetry breaking

Packing

Compressible to incompressible
Volume exclusion constraint

Courtesy of R. Bon, G. Theraulaz et al.

Sheep herds in the Mecantour
range, south-east of France.



Different types of phase transitions

Symmetry breaking

Packing

Continuum to networks

Emergent networks



S. Garnier

Expe: Perna, Granovskiy, Garnier, Nicolis, Labédan, Theraulaz, Fourcassié, Sumpter, PLoS CB 8 (2012), e1002592.

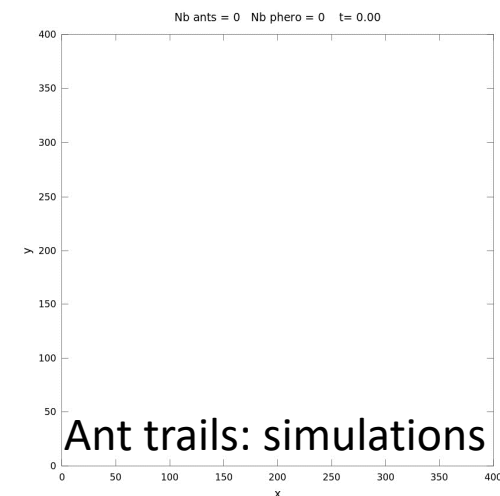
Simulations: Boissard, PD, Motsch, JTB 66 (2013) 1267



S. Motsch



Ant trails: ants enter arena from center and reach to the circular boundary

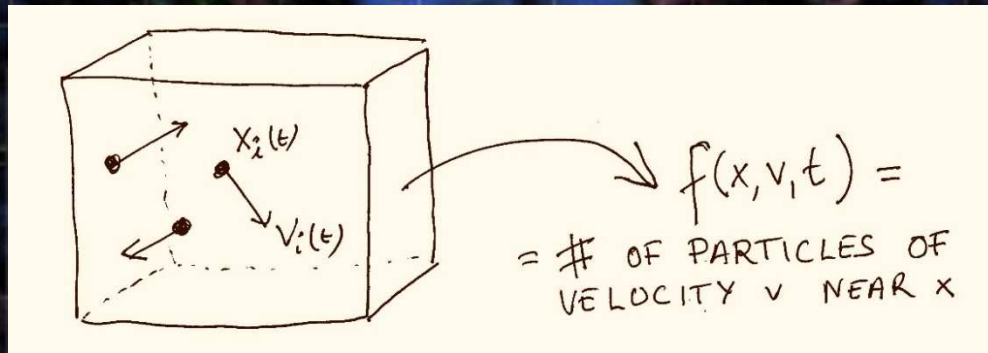


Ant trails: simulations

1st step: kinetic equation

Start with individual particles

Construct **Probability** $f=f(x,v,t)$



Equation for f requires influence of any given particle on the system be **very small**

Propagation of chaos

Propagation of chaos may be **untrue** for systems exhibiting emergence

Carlen, Chatelin, PD, Wennberg, Physica D
260 (2013) 90 & M3AS 23 (2013) 1339.



E. Carlen



B. Wennberg

2nd step: complexity reduction

Remove velocity variable by integration

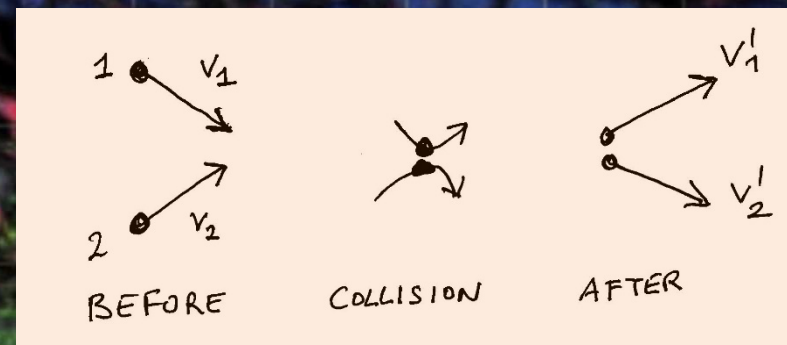
$$\rho(x,t) = \int f(x,v,t) dv \quad \text{PARTICLE DENSITY AT } x$$
$$\langle v \rangle(x,t) = \frac{1}{\rho(x,t)} \int f(x,v,t) v dv \quad \text{MEAN VELOCITY AT } x.$$

Macroscopic equations (for ρ and $\langle v \rangle$)
derived from **conservations**

In classical cases (gases):

CONSERVATION
OF
MOMENTUM

$$v'_1 + v'_2 = v_1 + v_2$$



No conservations for “exotic” particles

Ex. vehicles:
no momentum conservation



How to obtain macroscopic equations ?

“weaker” conservation:
“**generalized collision invariant**”

Application: **Vicsek**

Self propulsion + alignment + noise

Macroscopic model is

$$\partial_t \rho + c_1 \nabla_x (\rho u) = 0$$

$$\rho (\partial_t u + c_2 (u \cdot \nabla_x) u) + P_{u^\perp} \nabla_x \rho = 0$$

$$|u| = 1$$



S. Motsch

PD, Motsch, M3AS 18
Suppl (2008) 1193

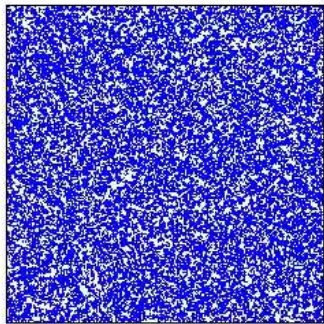
Self-Organized Hydrodynamics (SOH)

Vicsek (micro) vs SOH (macro)

Micro model (Vicsek)

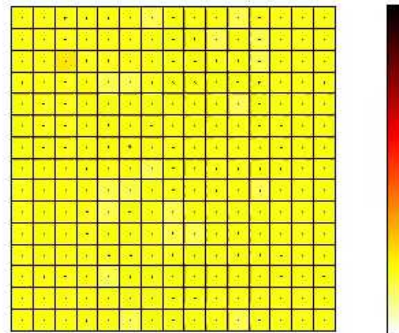
Self propulsion + alignment + noise

Particles at $t = 0.00$



Particles

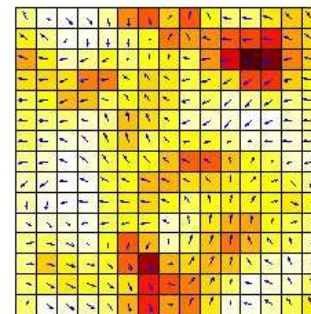
Density and velocity at $t = 0.00$



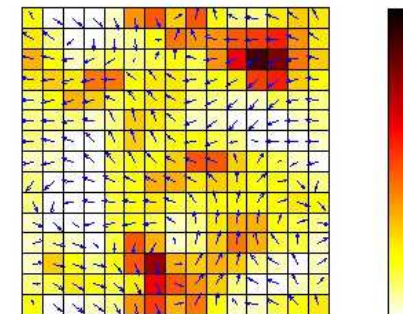
Density (color)
Velocity (arrows)

Micro (Vicsek) Macro (SOH)

Micro at $t = 20.00$



Macro at $t = 20.00$



Density (color)
Velocity (arrows)

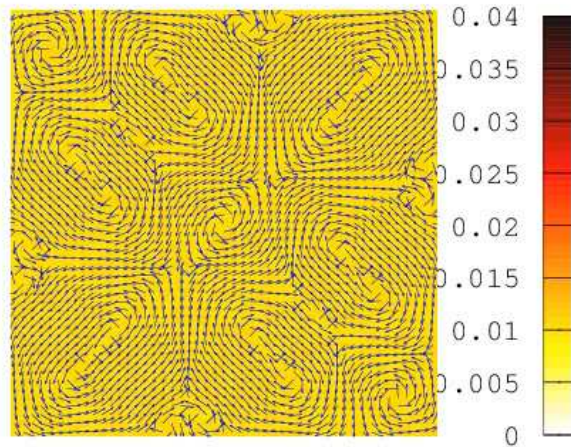


Simulations by S. Motsch

Vicsek (micro) vs SOH (macro)

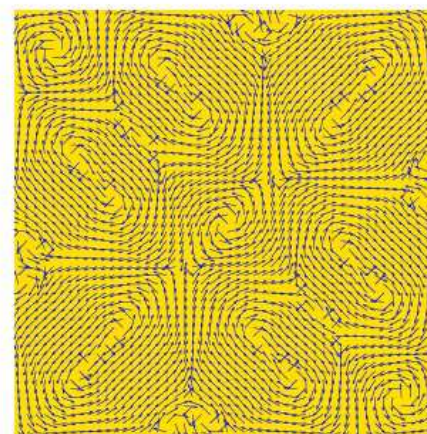
Micro (Vicsek)
Self propulsion + alignment
+ noise + repulsion

Micro at $t = 0.00$



Macro (SOH)

Macro at $t = 0.00$



Density (color)
Velocity (arrows)

Simulations by G. Dimarco and N. Mac



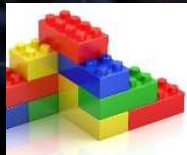
G. Dimarco



Cells: 2D spheres



Fibers: line segments



Cell-cell volume exclusion



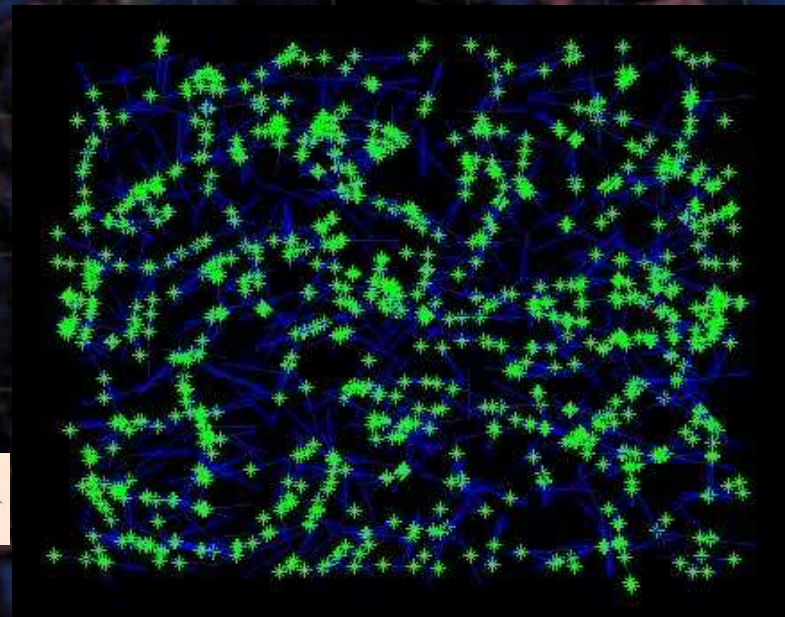
Cell-fiber repulsion



Fiber-fiber cross-linking



Fiber-fiber alignment



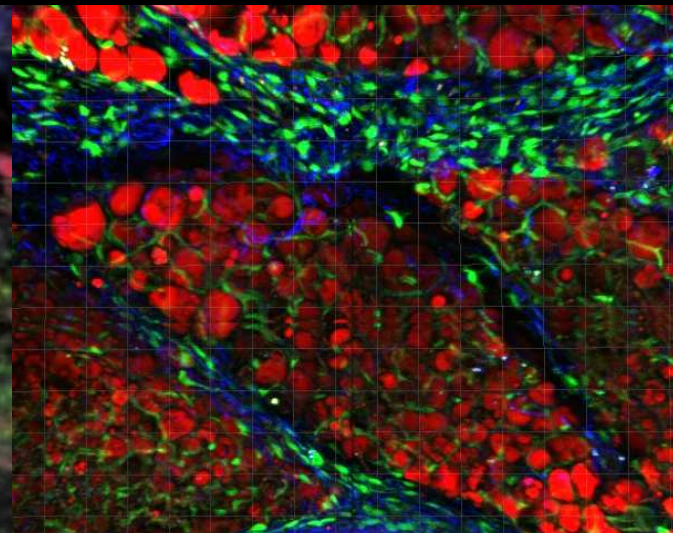
D. Peurichard, F. Delebecque, A.
Lorsignol, C. Barreau, J.
Rouquette, X. Descombes, L.
Casteilla, PD, sub.



L. Casteilla



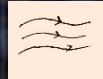
D. Peurichard



Blood capillary formation



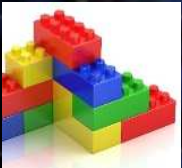
Blood: fluid



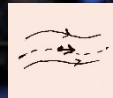
Oxygen (O₂): point particles



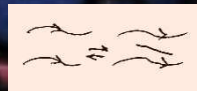
Capillaries: line segments



Oxygen transport by blood



Capillary creation/deletion
In response to blood/O₂



Blood/O₂ transport
enhanced by capillaries

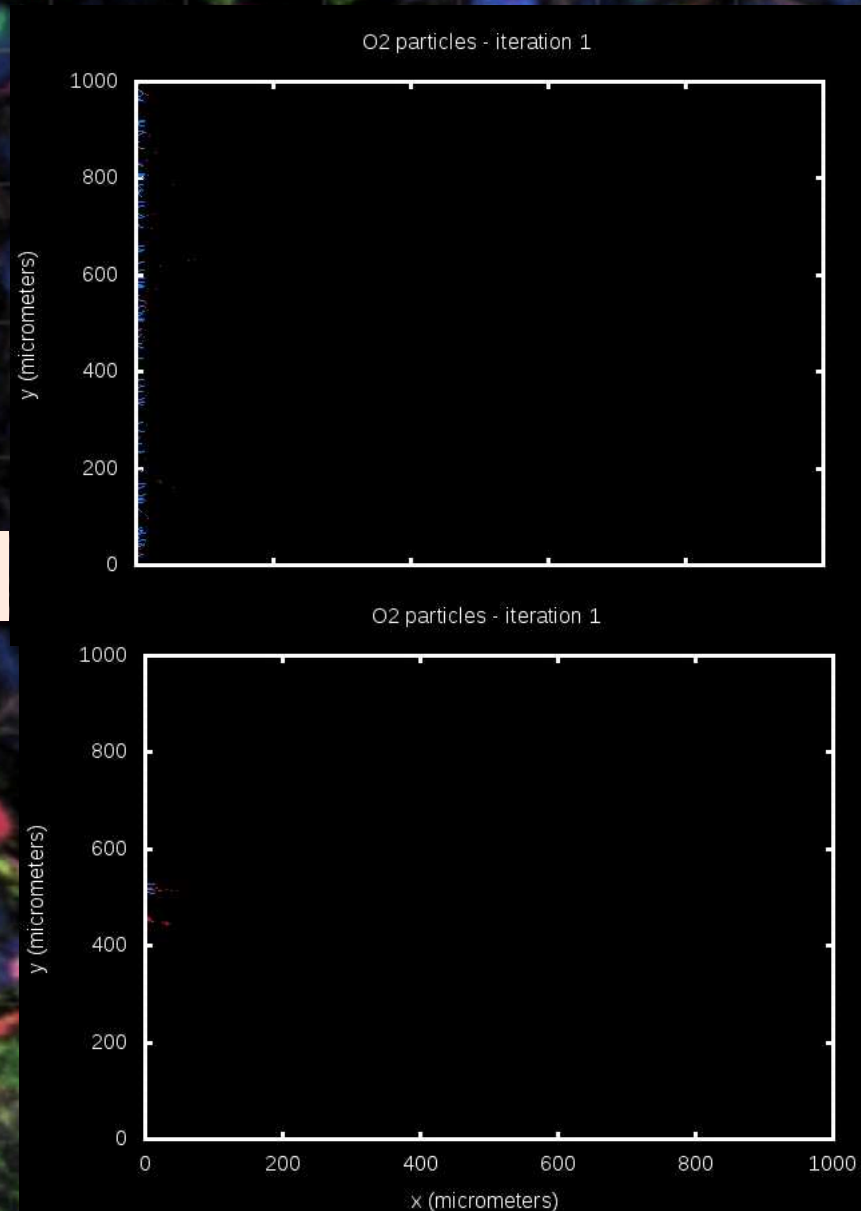


B. Aymard

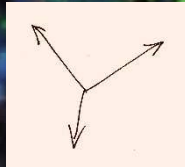
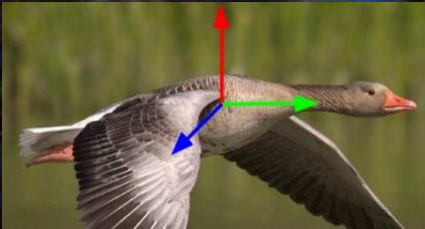


F. Plouraboué

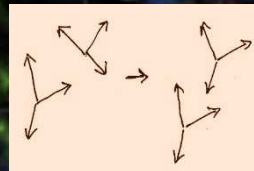
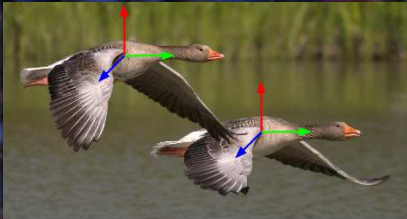
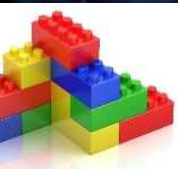
B. Aymard, A. Lorsignol,
L. Casteilla, P. Kennel, F.
Plouraboué, PD, in
preparation



Birds, fish: frames



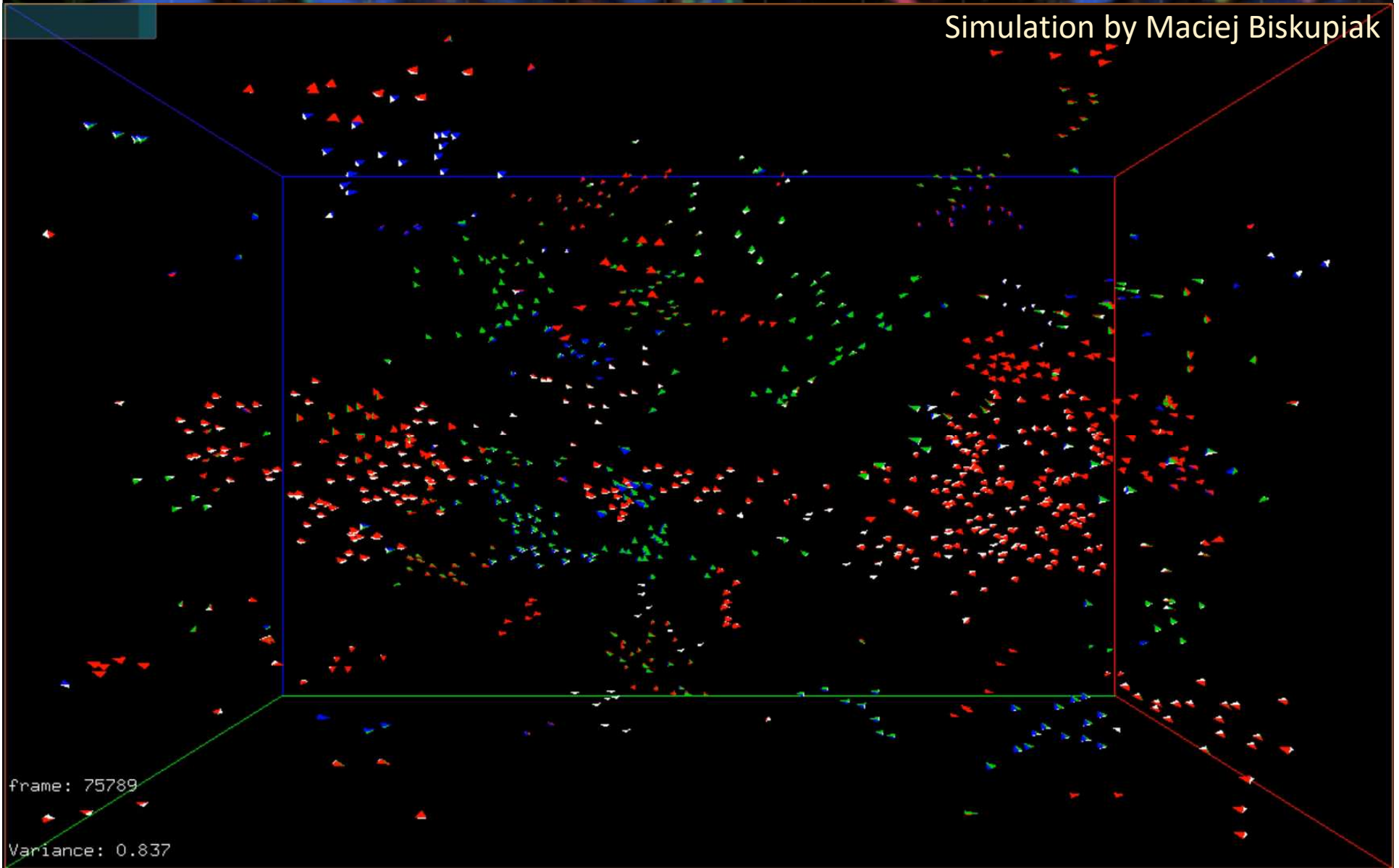
Frame alignment



Simulation by Maciej Biskupiak

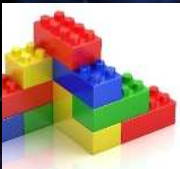
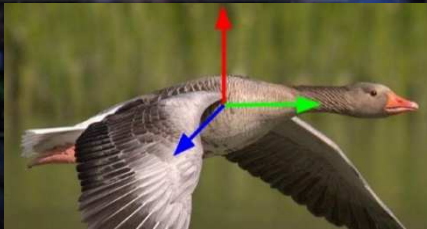
frame: 75789

Variance: 0.837

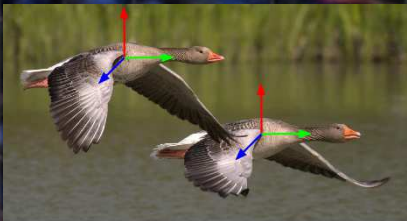




Birds, fish: frames



Frame alignment



Macroscopic model:

$$\begin{aligned} \partial_t \rho + c_1 \nabla_x \cdot (\rho \Lambda \mathbf{e}_1) &= 0, \\ \rho \left(\partial_t \Lambda + c_2 ((\Lambda \mathbf{e}_1) \cdot \nabla_x) \Lambda \right) \\ &+ [(\Lambda \mathbf{e}_1) \times (c_3 \nabla_x \rho + c_4 \rho \mathbf{r}_x(\Lambda))] \\ &+ c_4 \rho \delta_x(\Lambda) \Lambda \mathbf{e}_1 \Big]_{\times} \Lambda = 0. \end{aligned}$$



A. Frouvelle



S. Merino-Aceituno

PD, A. Frouvelle, S.
Merino-Aceituno,
arXiv:1605.03509
to appear in M3AS



Emergence: Property of systems that develop patterns on scales larger than those of their individual components



Emergent systems are **important** in science and engineering



Emergence is a **phase transition**: a brutal change of the system's properties in response to small parameter changes

$$f(x, v, t)$$

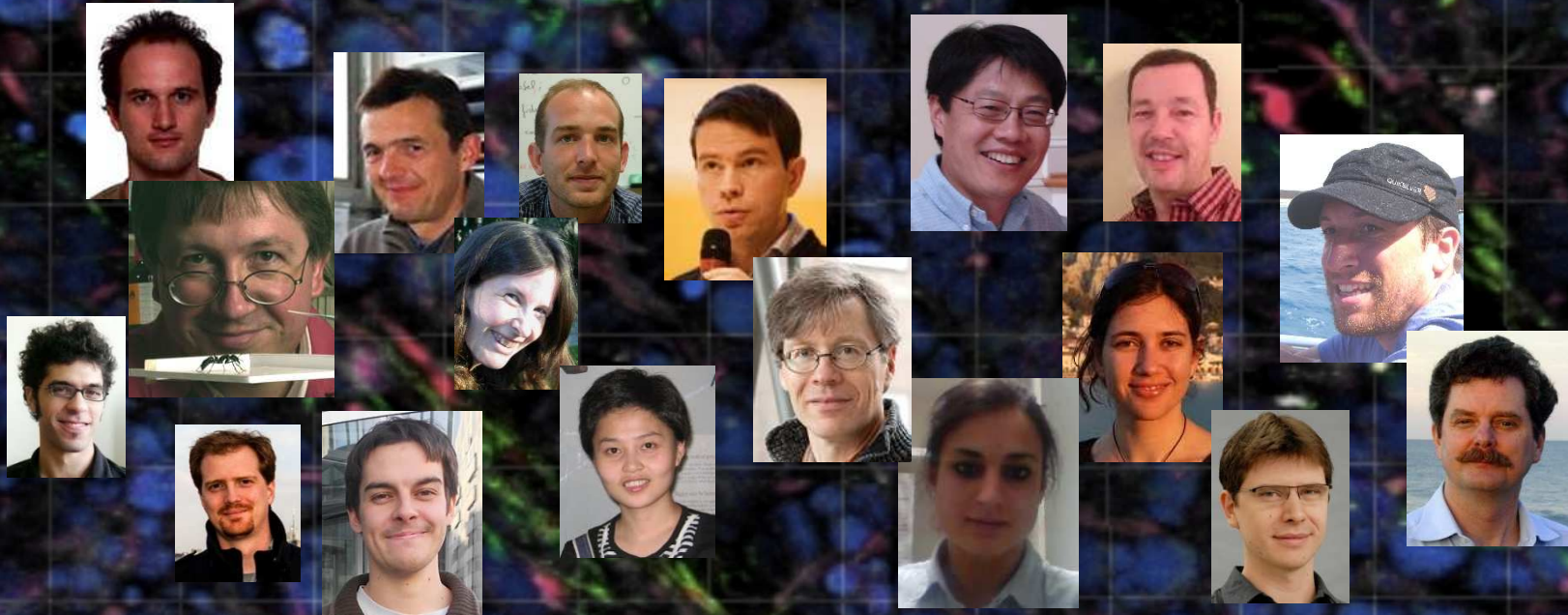
Kinetic theory is a method of choice to derive models of emergent systems in line of **Hilbert's** 6th problem



But emergence requires developing concepts **beyond** the state of the art

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Special thanks to all my
collaborators and students



and to the many missing here !!

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Special thanks to my sponsors

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Since I'm in UK



When I was in France

