

The auxiliary region method for coupling PDE and Brownian- based dynamics for reaction-diffusion systems



Cameron Smith (Cohort 3)

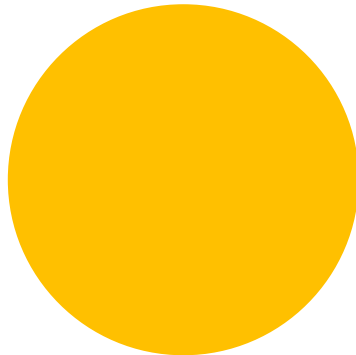
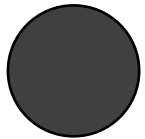
SAMBa Conference 2019
08/07/19

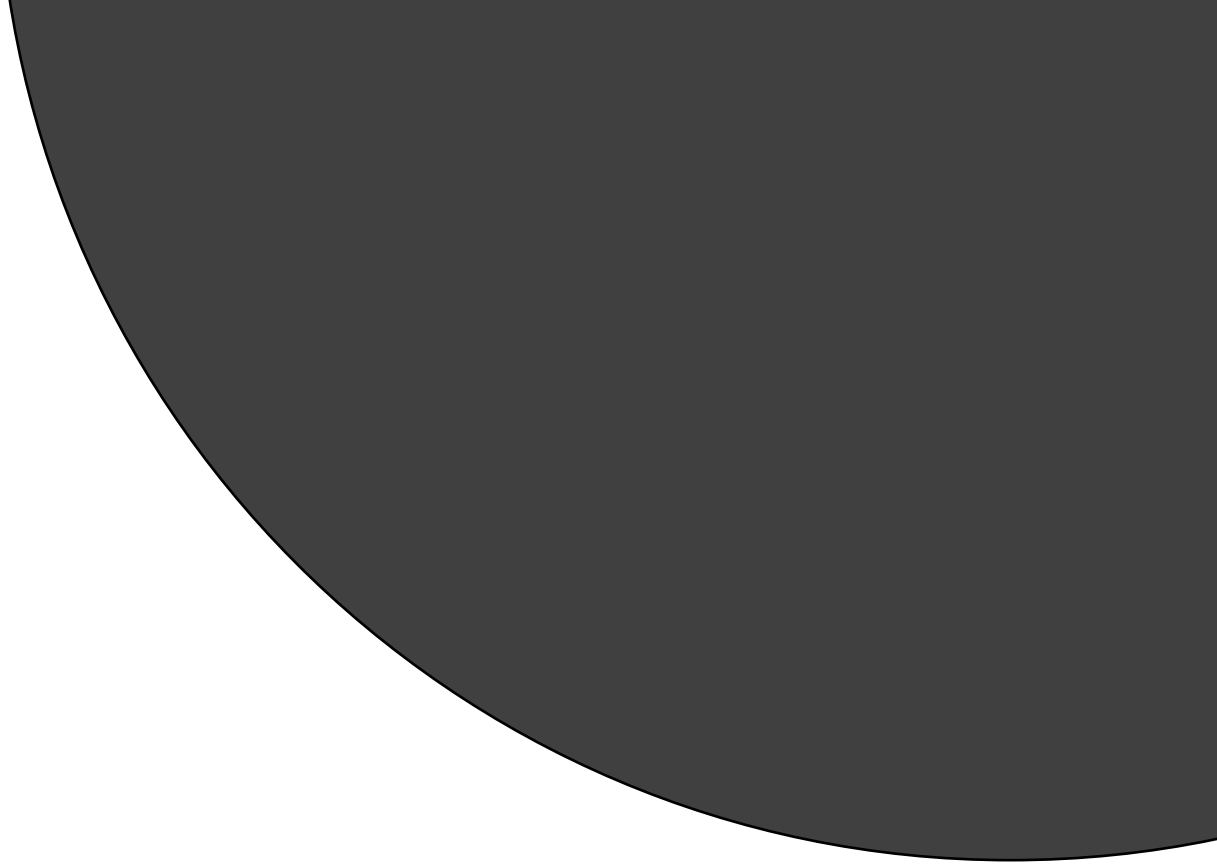
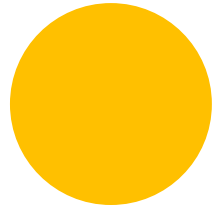
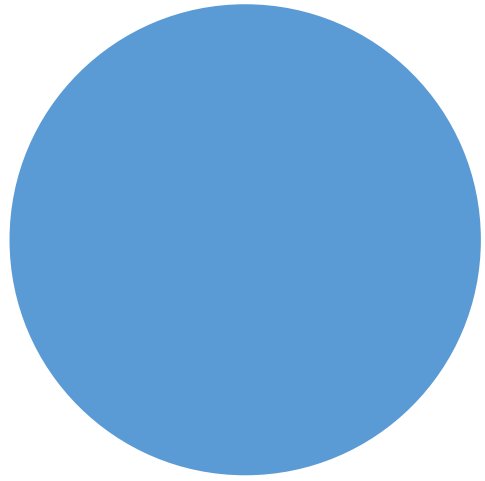
Outline

Reaction-diffusion systems

Spatially extended hybrid methods

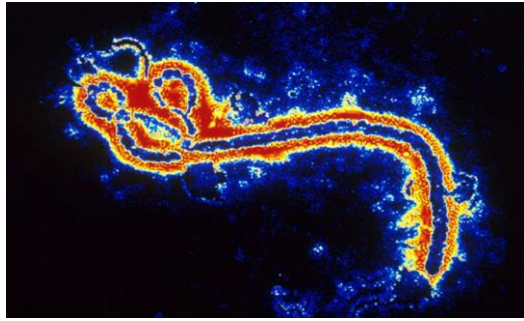
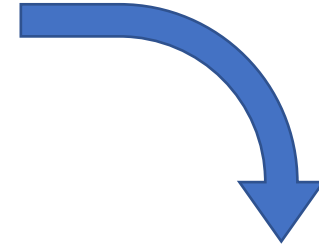
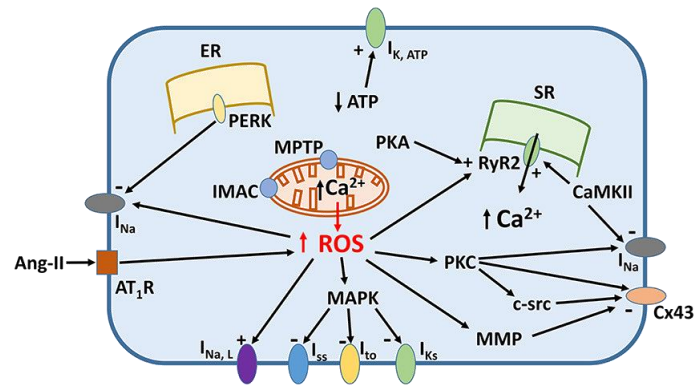
The auxiliary region method (ARM)



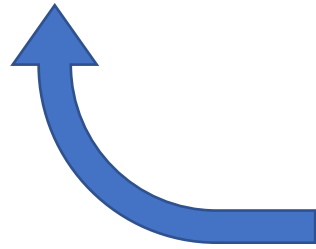
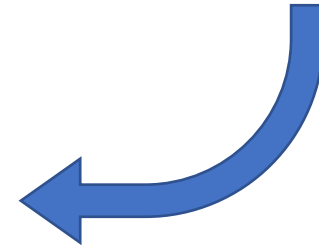
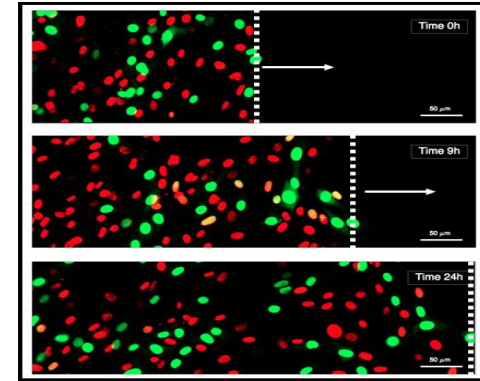


Reaction-diffusion systems

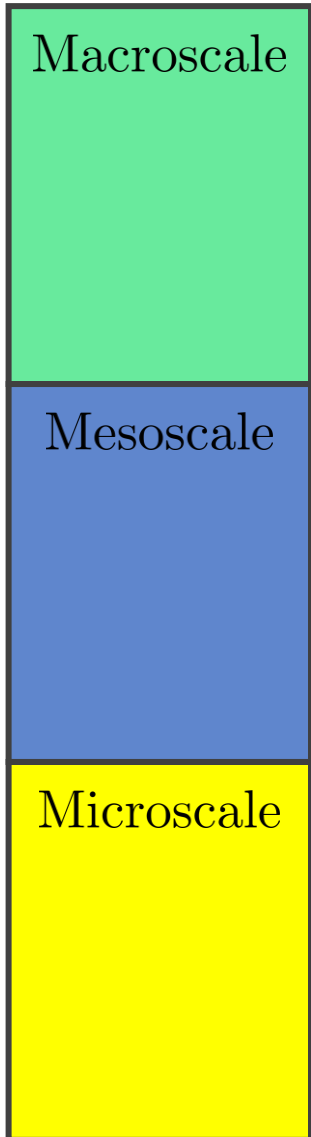




Reaction-diffusion



Modelling



Modelling

Macroscale

Partial differential equation

$$\frac{\partial u}{\partial t} = D \nabla^2 u + \mathcal{R}[u]$$

Diffusion

Reactions

Mesoscale

Microscale

✓ Many numerical techniques

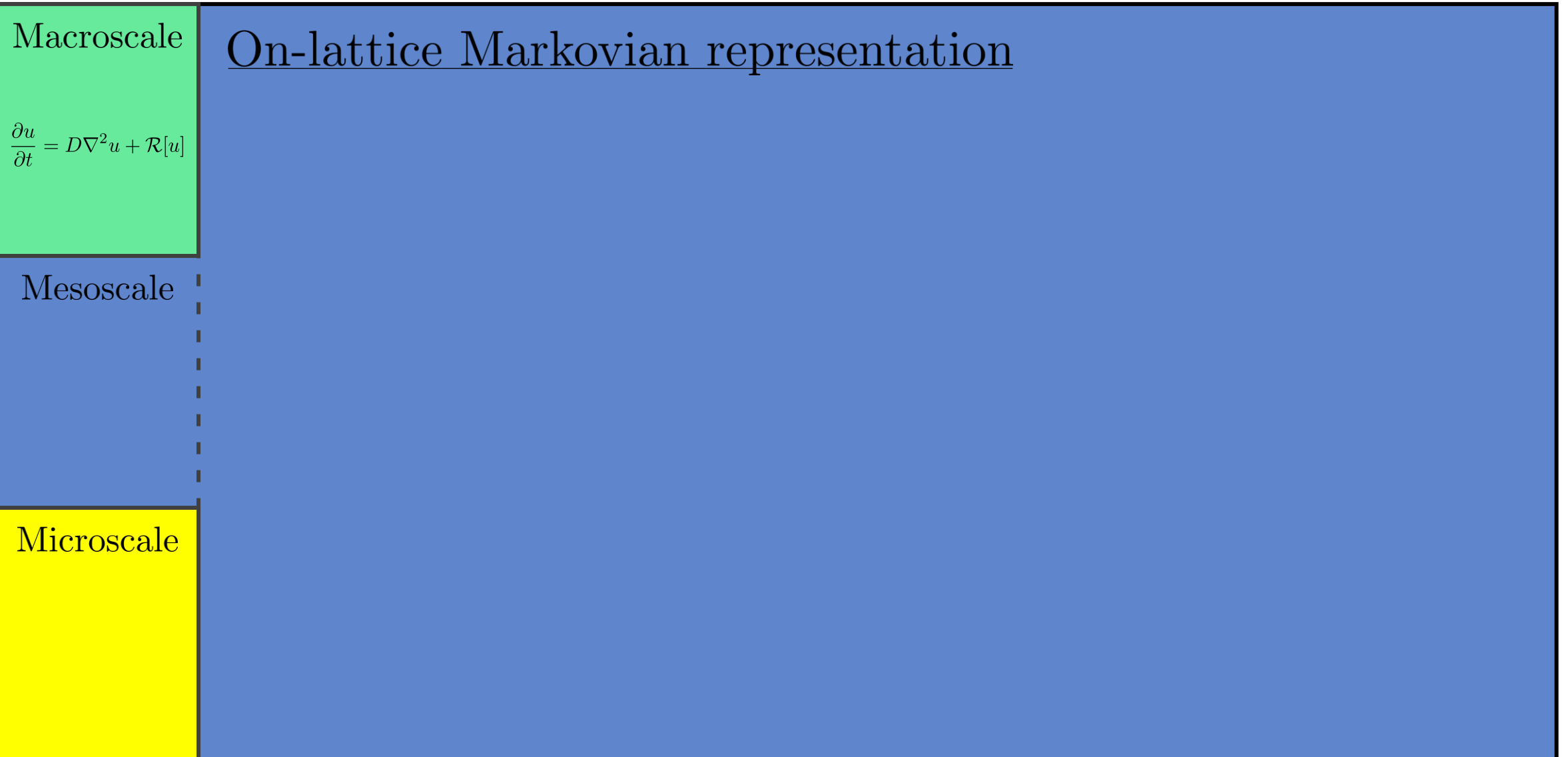
✗ Unsuitable for low copy numbers

✓ Amenable to analysis

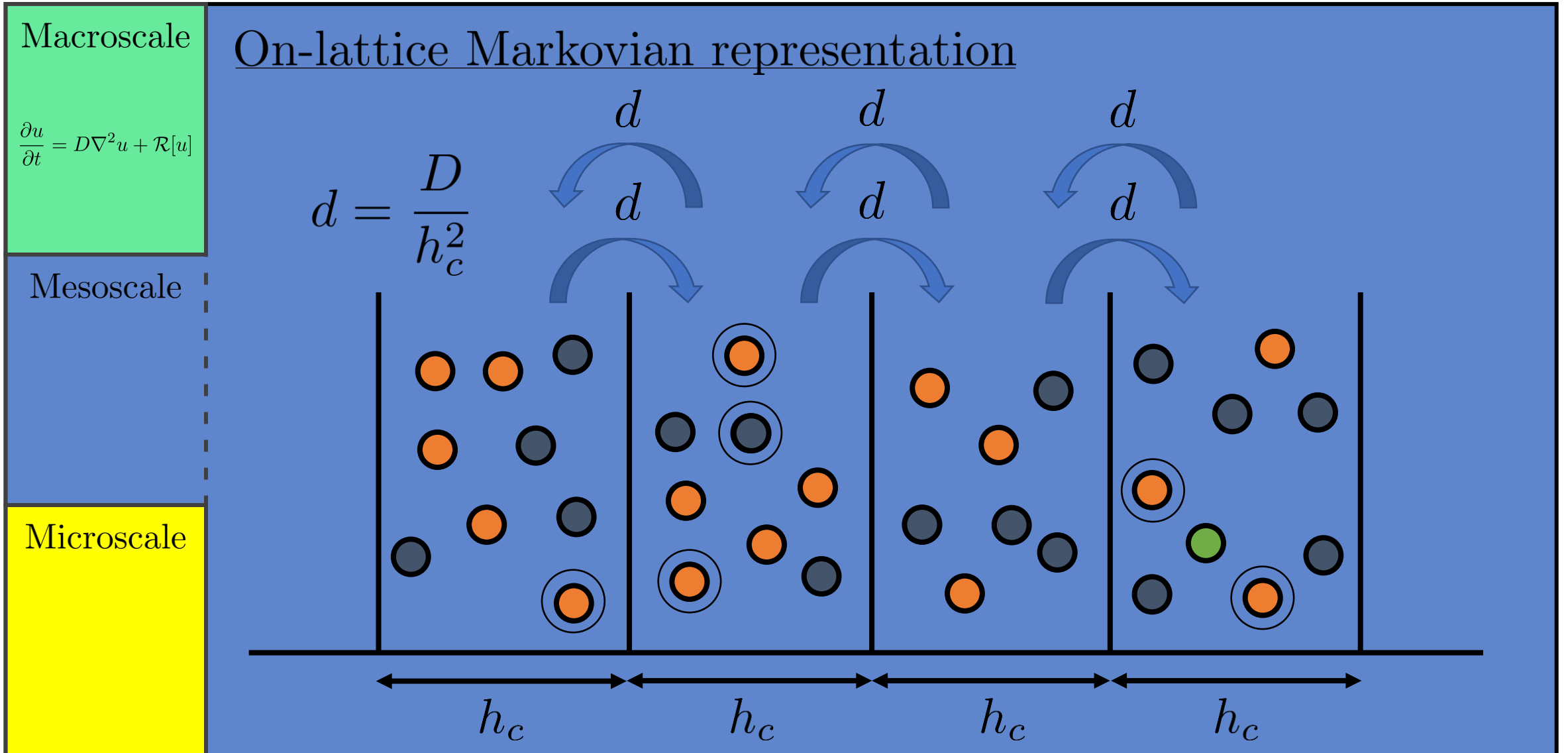
✗ Can be inaccurate

✓ Fast to simulate

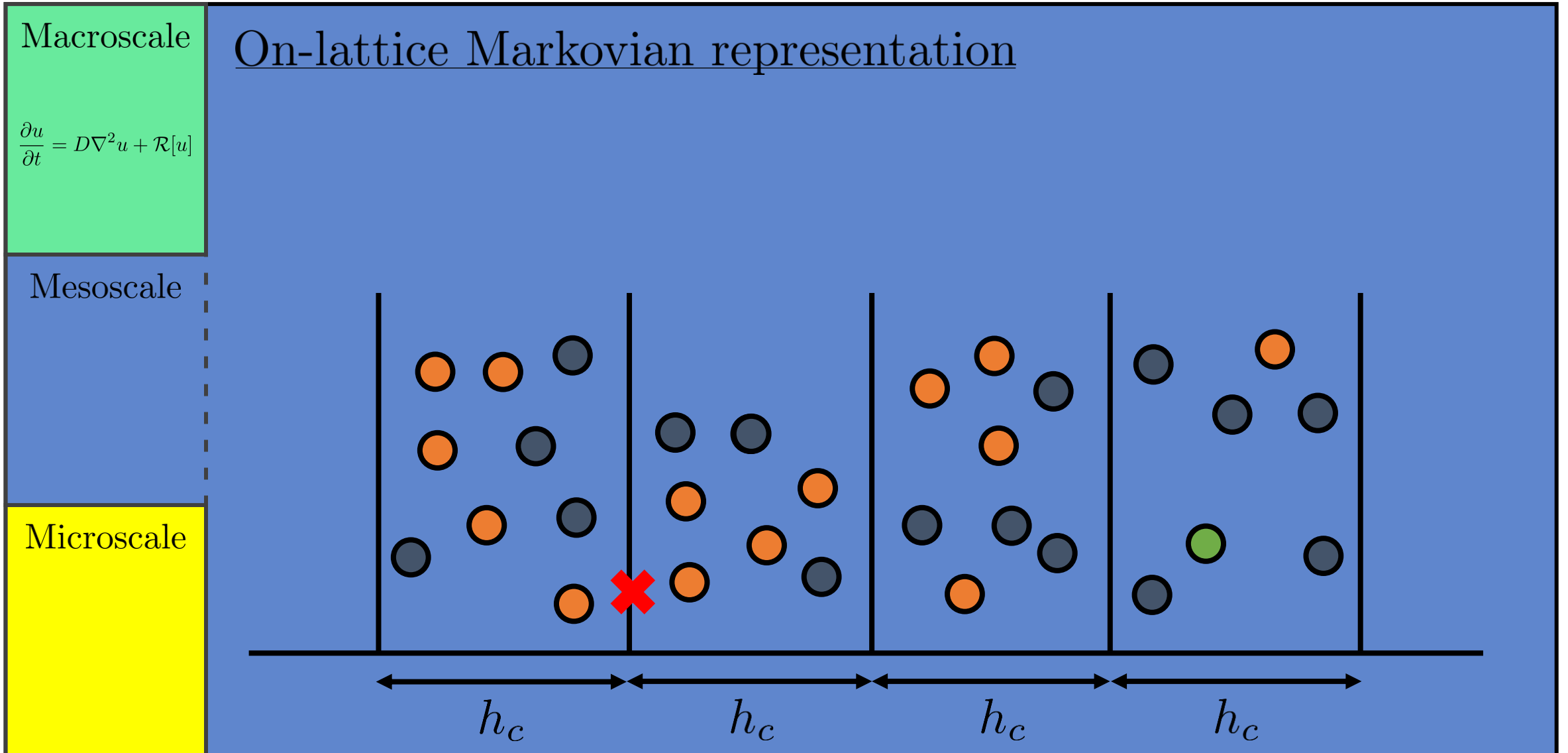
Modelling



Modelling



Modelling



Modelling

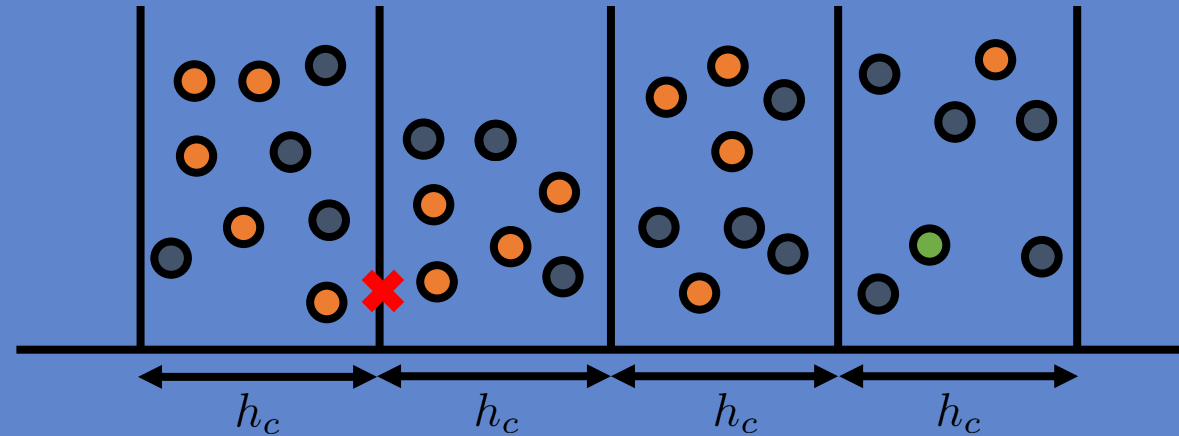
Macroscale

$$\frac{\partial u}{\partial t} = D\nabla^2 u + \mathcal{R}[u]$$

Mesoscale

Microscale

On-lattice Markovian representation



Many numerical techniques

Slow for large numbers of particles

Stochastic information

Lose particle "locations"

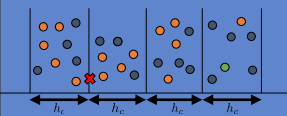
Reactions are simple to implement

Modelling

Macroscale

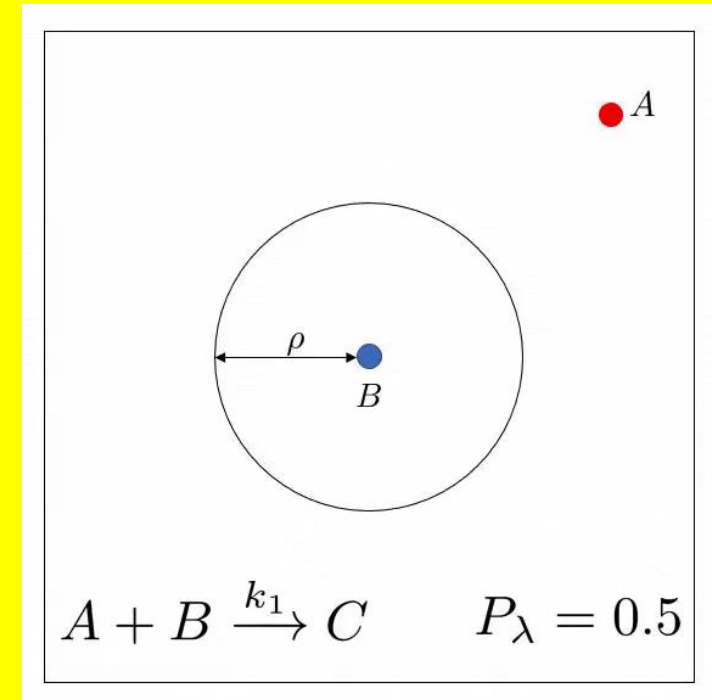
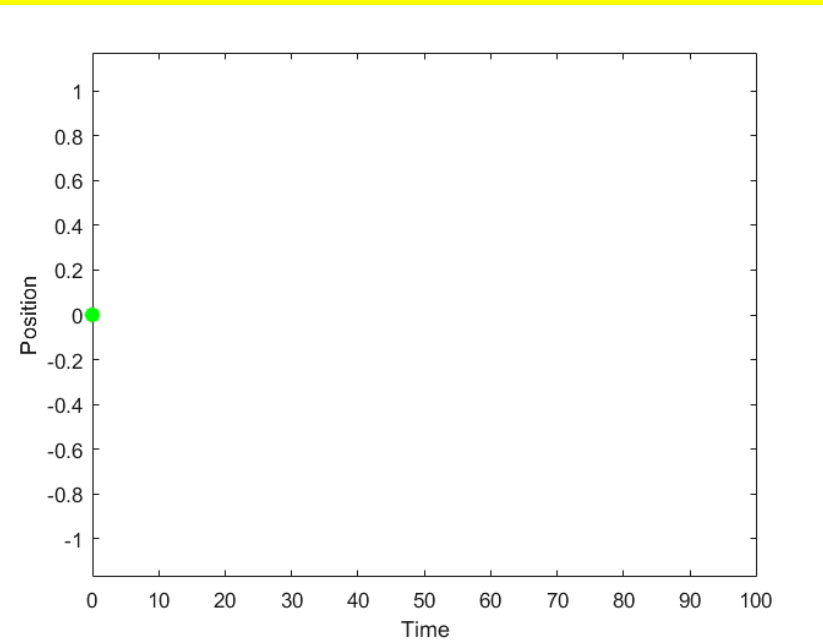
$$\frac{\partial u}{\partial t} = D\nabla^2 u + \mathcal{R}[u]$$

Mesoscale



Microscale

Off-lattice Brownian-based

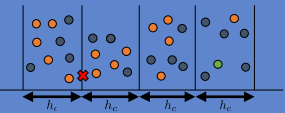


Modelling

Macroscale

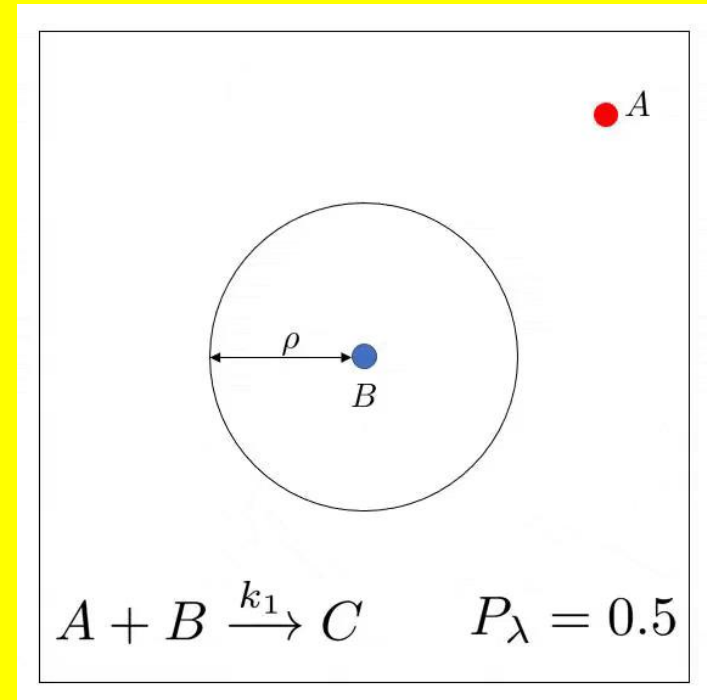
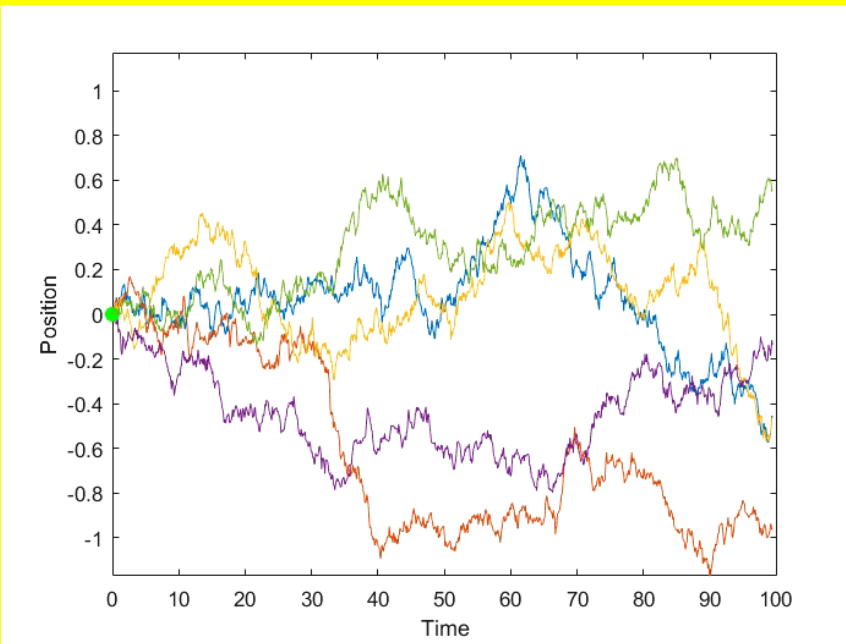
$$\frac{\partial u}{\partial t} = D\nabla^2 u + \mathcal{R}[u]$$

Mesoscale



Microscale

Off-lattice Brownian-based

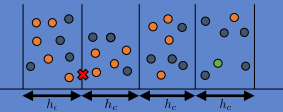


Modelling

Macroscale

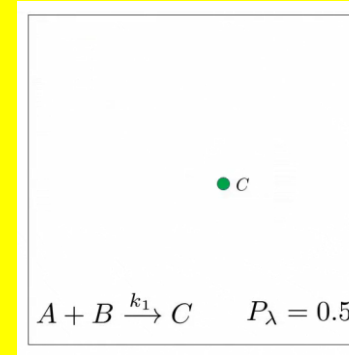
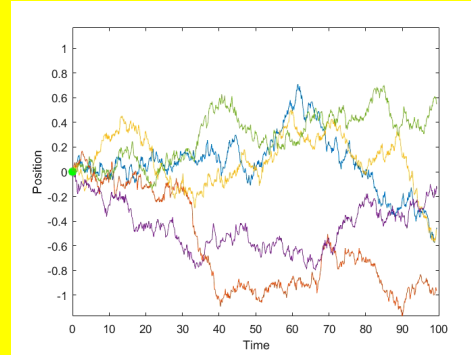
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Mesoscale



Microscale

Off-lattice Brownian-based



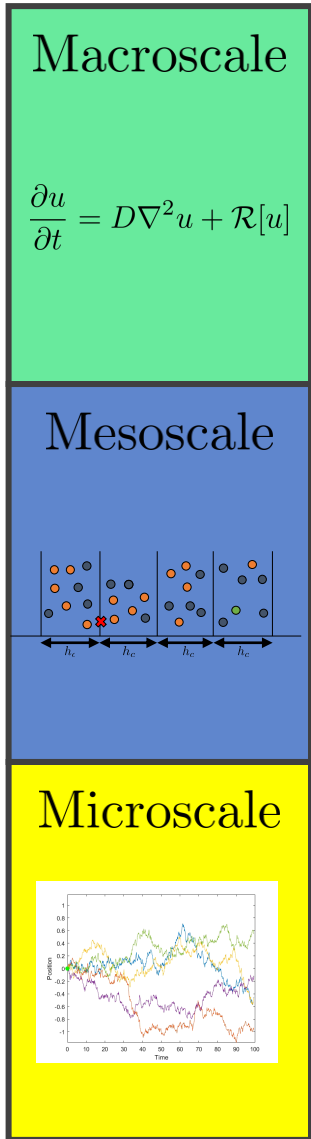
Individual level behaviour

Slow to implement

Stochastic information

May require pairwise distances

Modelling

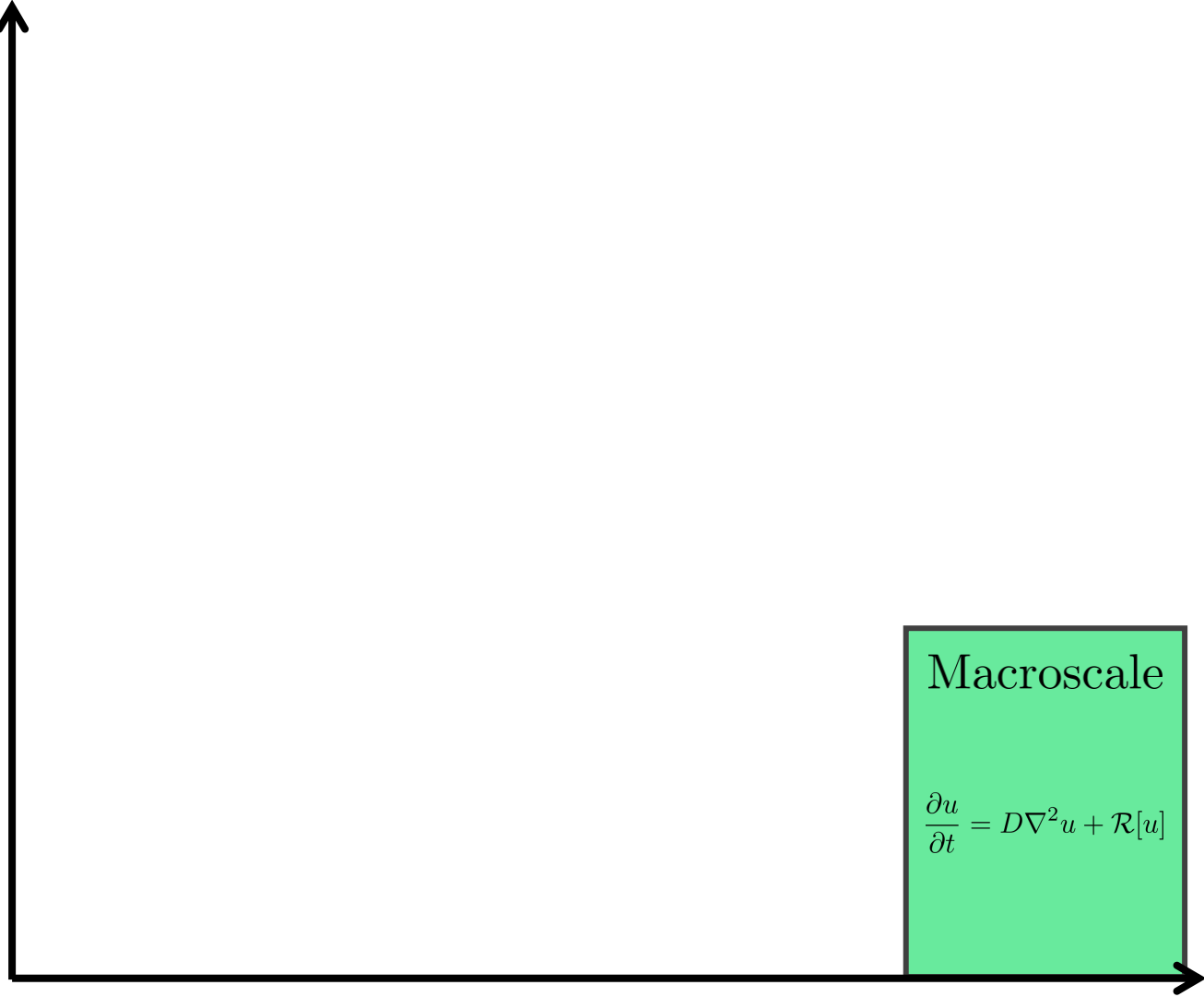


Accuracy ↑

Efficiency →

Modelling

Accuracy



Macroscale

$$\frac{\partial u}{\partial t} = D\nabla^2 u + \mathcal{R}[u]$$

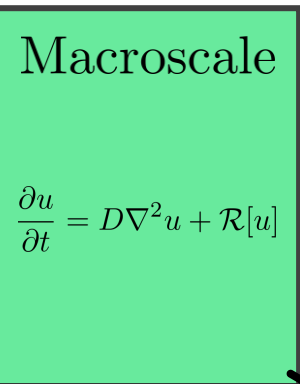
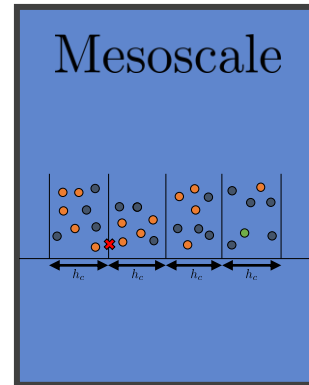
Efficiency

Mesoscale

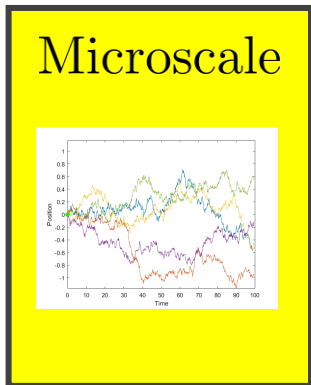
Microscale

Modelling

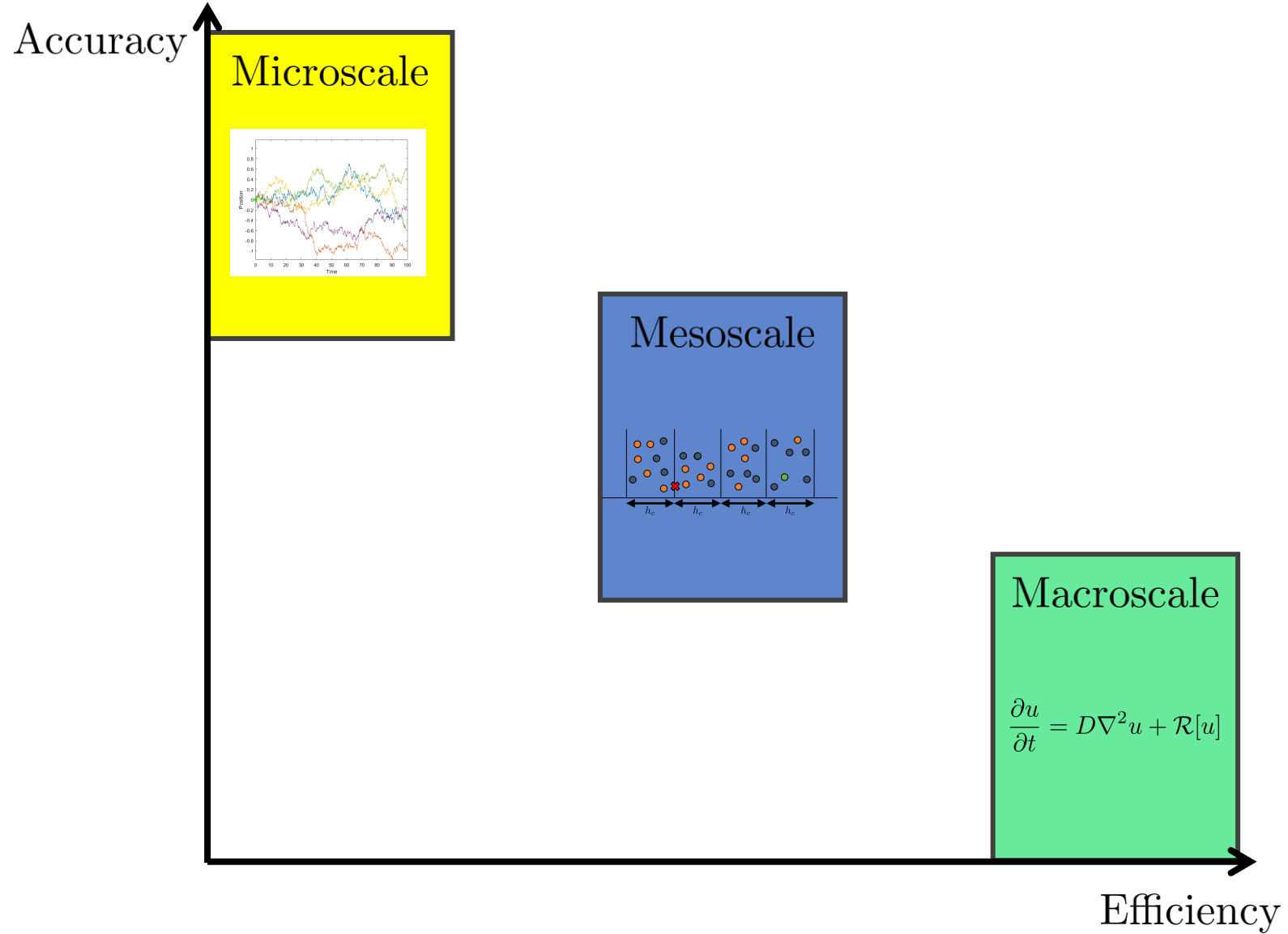
Accuracy

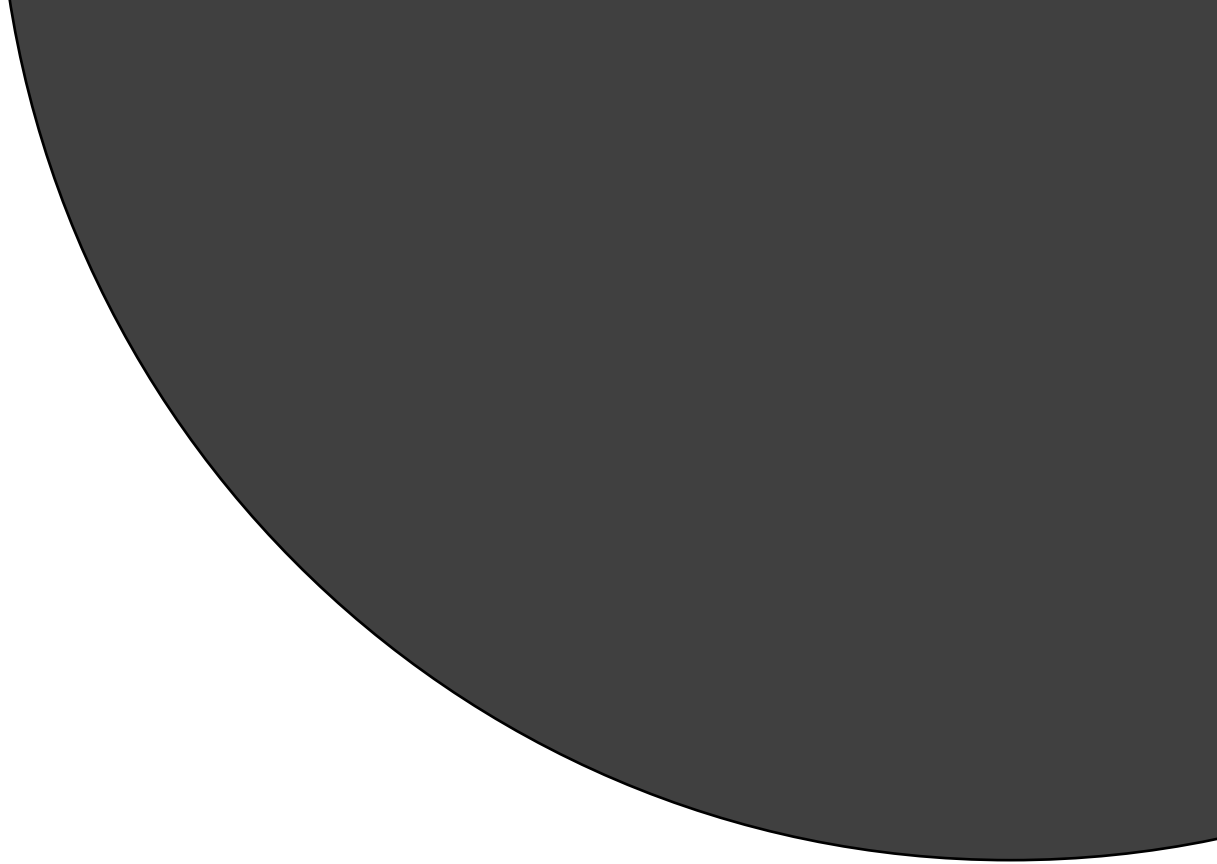
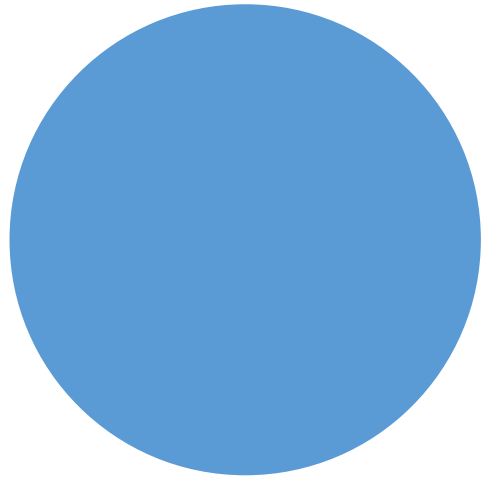


Efficiency



Modelling





Spatially extended
hybrid methods





“

... spatially extended hybrid methods employ different modelling paradigms at different scales in order to compliment the strengths and negate the weaknesses of each.

”

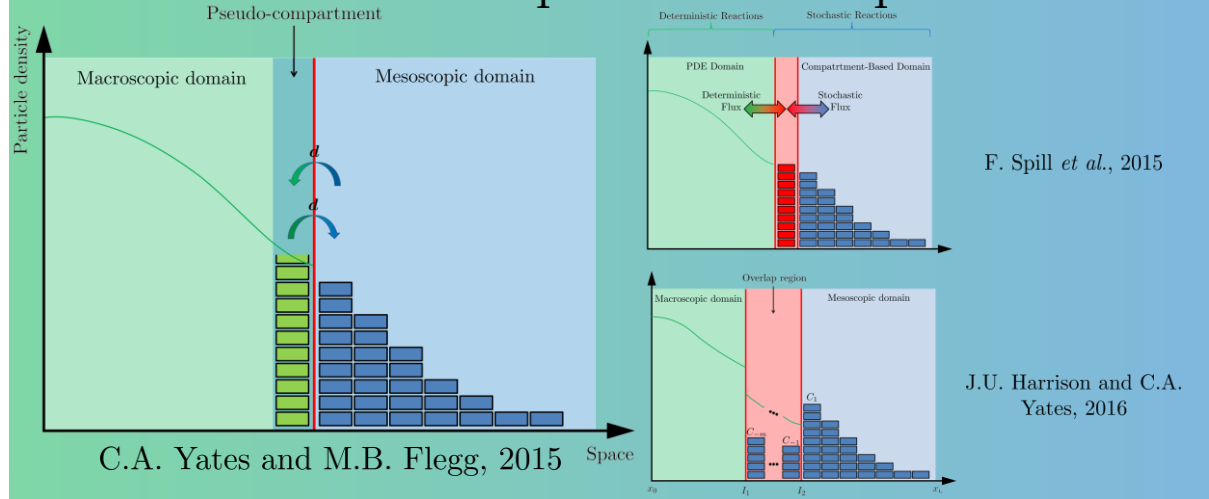


C.A. Smith and C.A. Yates, 2018

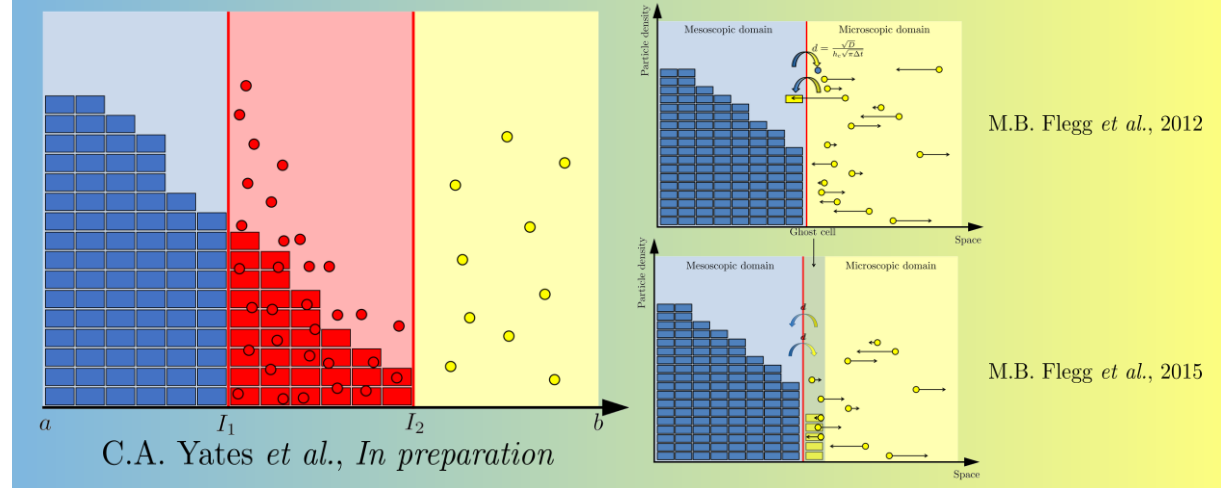
Spatially extended hybrid methods: a review

Many, many examples!

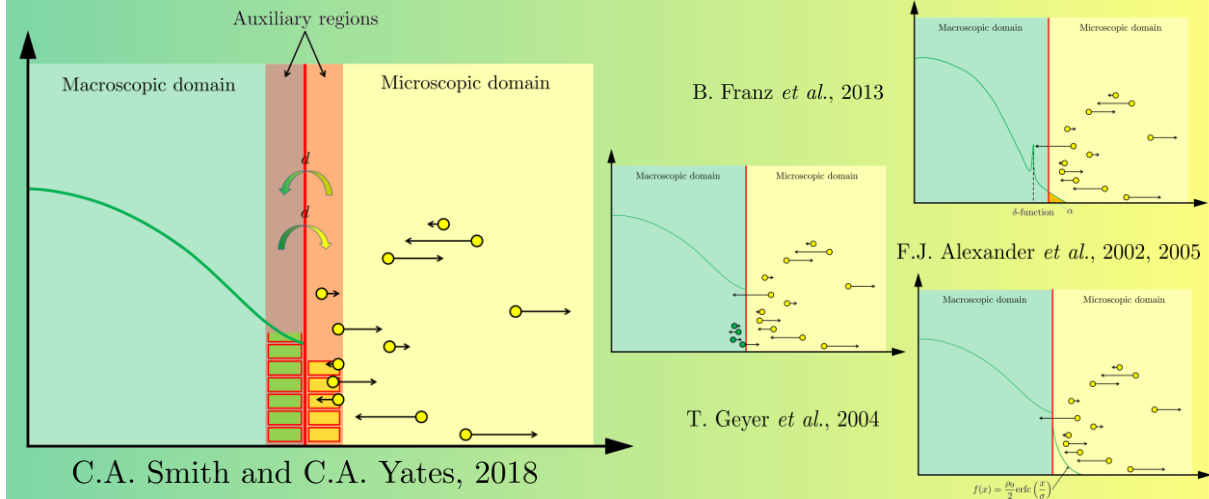
Macroscopic-to-mesoscopic



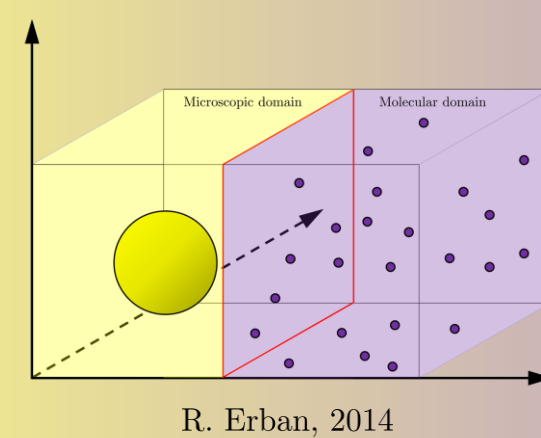
Mesoscopic-to-microscopic

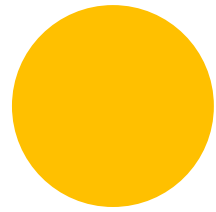
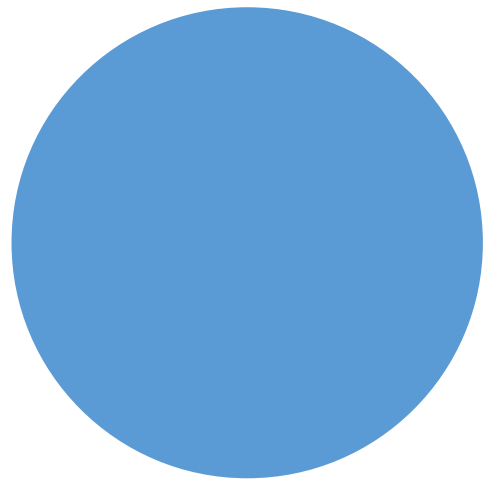


Macroscopic-to-microscopic



Microscopic-to-molecular

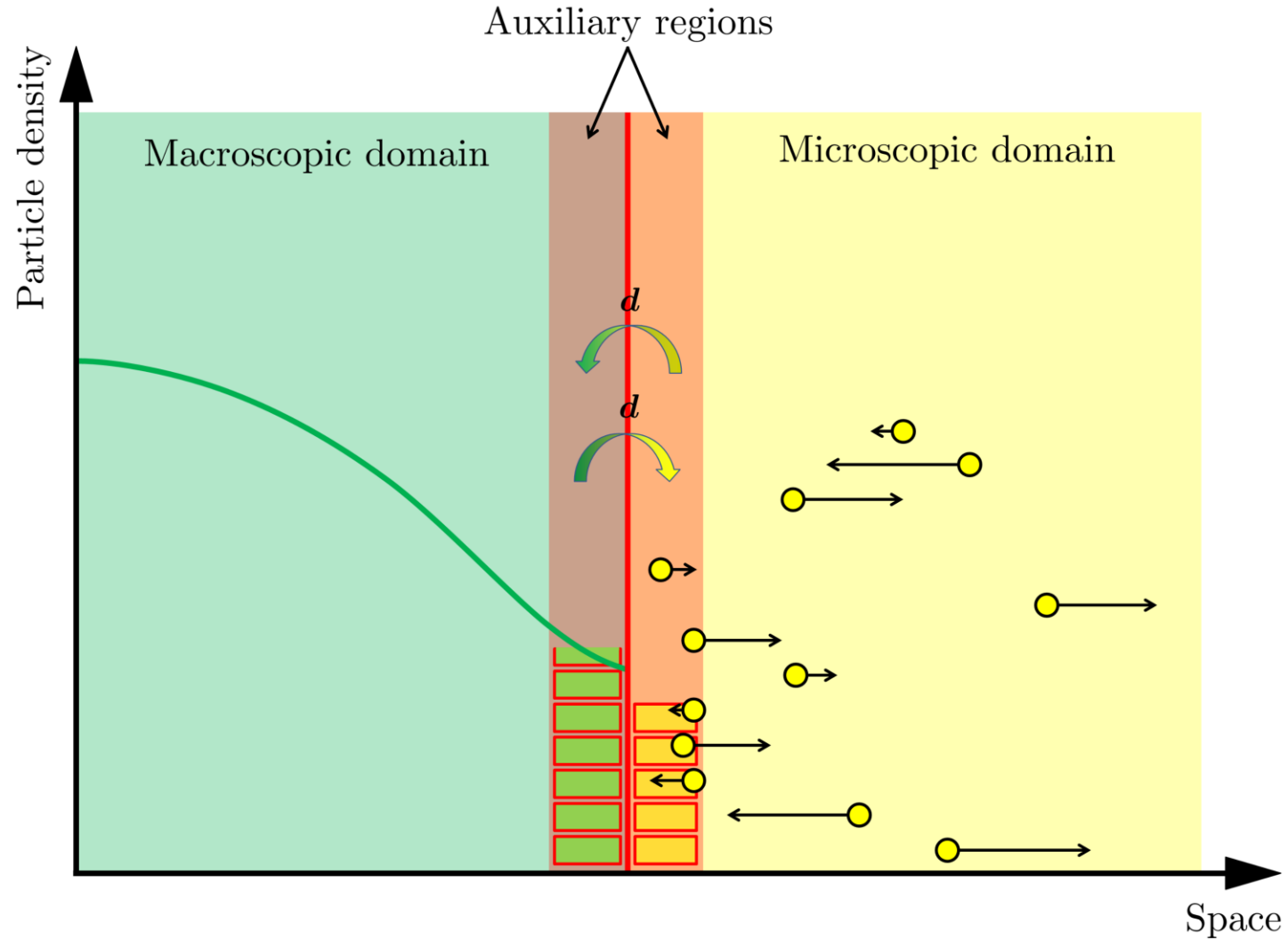




The auxiliary region method



The auxiliary region method



Basic algorithm

Let the PDE/Brownian update step be Δt , and t_Δ be the next PDE/Brownian update time.

- 1) Find the time until the next event within the auxiliary regions occurs. Call this t_a .
- 2) If this is less than the time until the next PDE/Brownian update (i.e. if $t_a < t_\Delta$), find the corresponding event and enact it.
- 3) Otherwise (i.e. if $t_\Delta < t_a$), evolve the PDE and Brownian subdomains.
- 4) Update time and return to step 1.

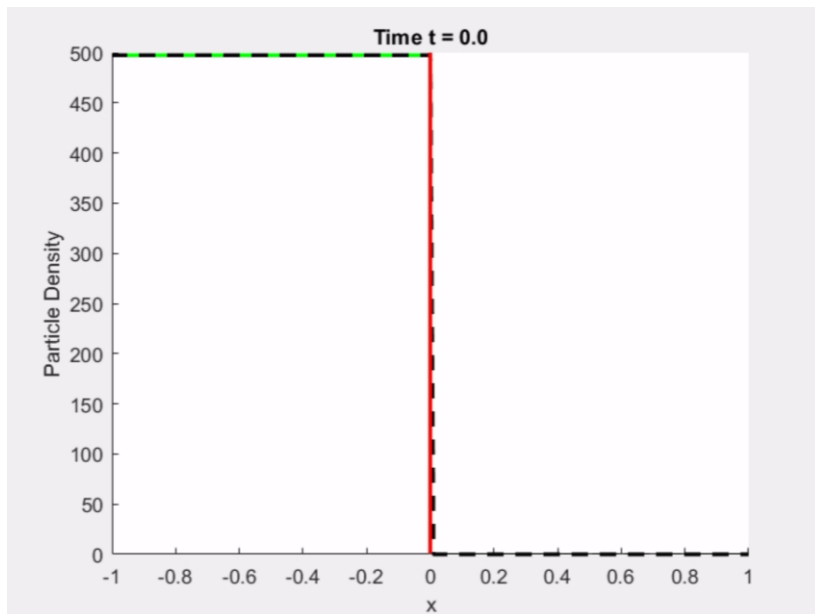
Results

Pure diffusion

$$\frac{\partial u}{\partial t} = D\nabla^2 u$$

$$\left. \frac{\partial u}{\partial x} \right|_{x=-1} = 0 \text{ and } \left. \frac{\partial u}{\partial x} \right|_{x=1} = 0$$

$$u(x, 0) = \begin{cases} 500 & x \in [-1, 0) \\ 0 & x \in [0, 1] \end{cases}$$

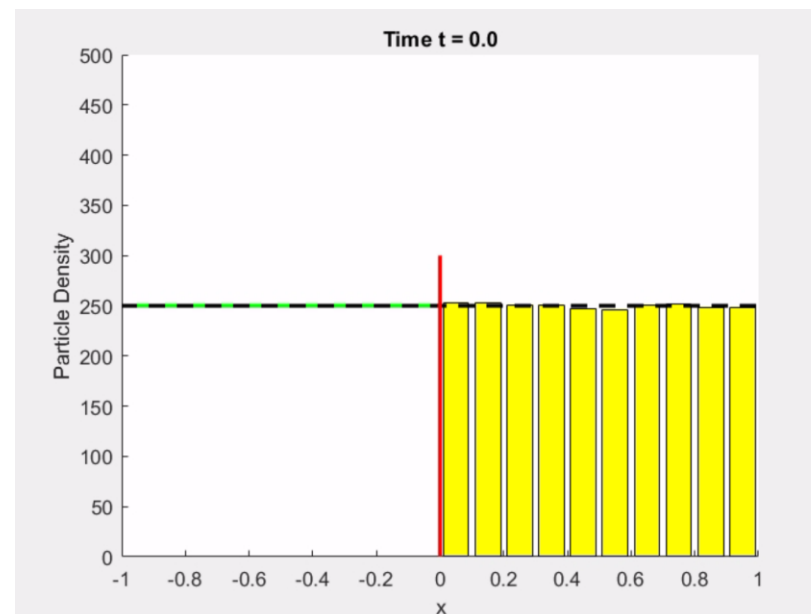


Morphogen gradient

$$\frac{\partial u}{\partial t} = D\nabla^2 u - \mu u$$

$$\left. \frac{\partial u}{\partial x} \right|_{x=-1} = -\lambda \text{ and } \left. \frac{\partial u}{\partial x} \right|_{x=1} = 0$$

$$u(x, 0) = 250 \quad \forall x \in [-1, 1]$$



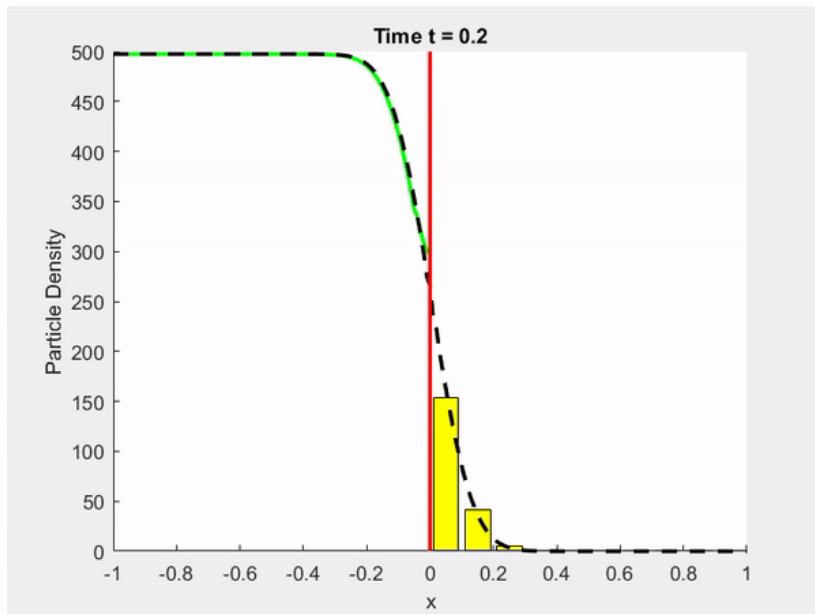
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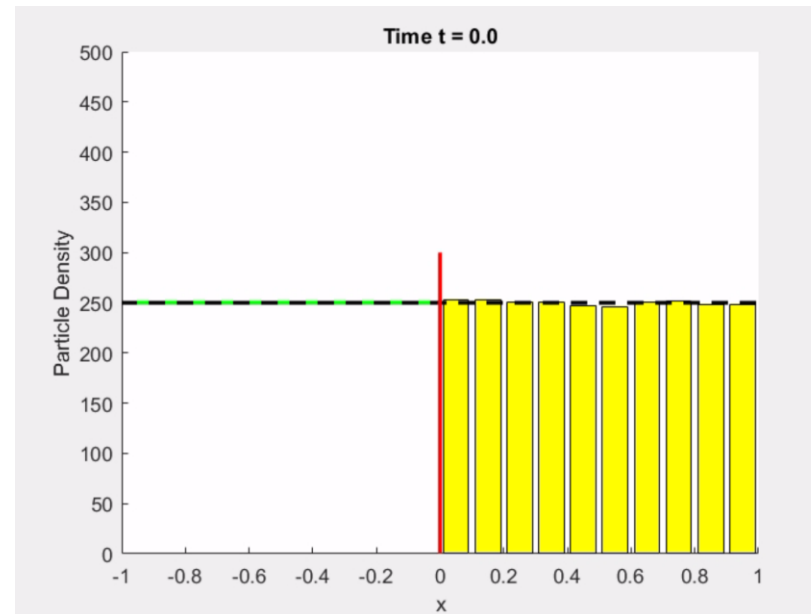


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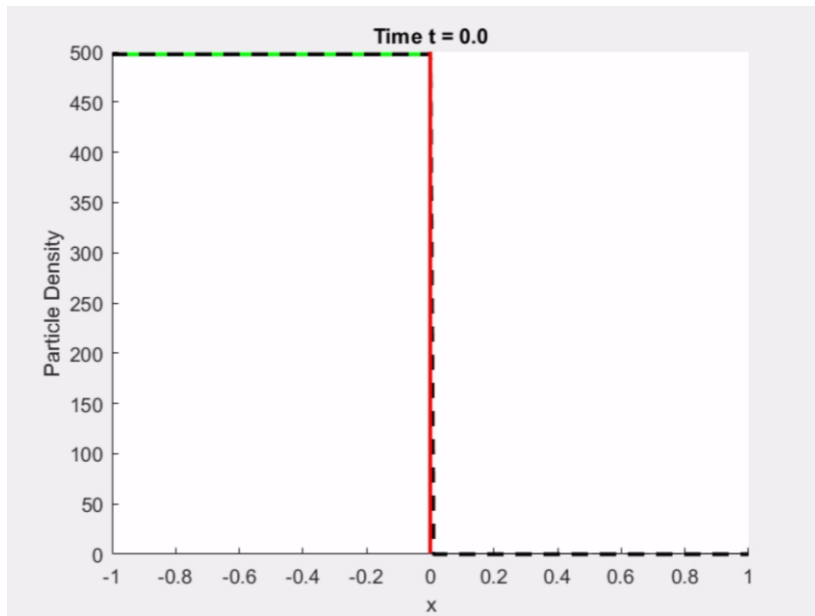
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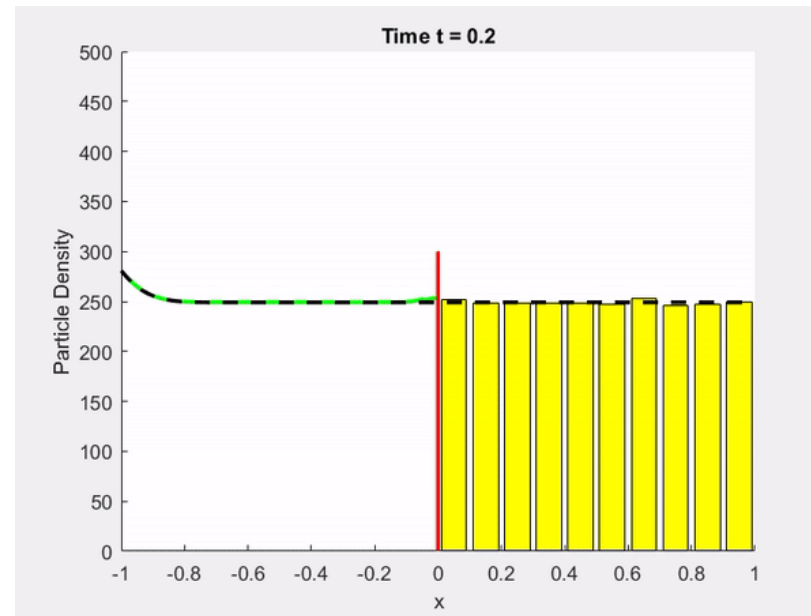


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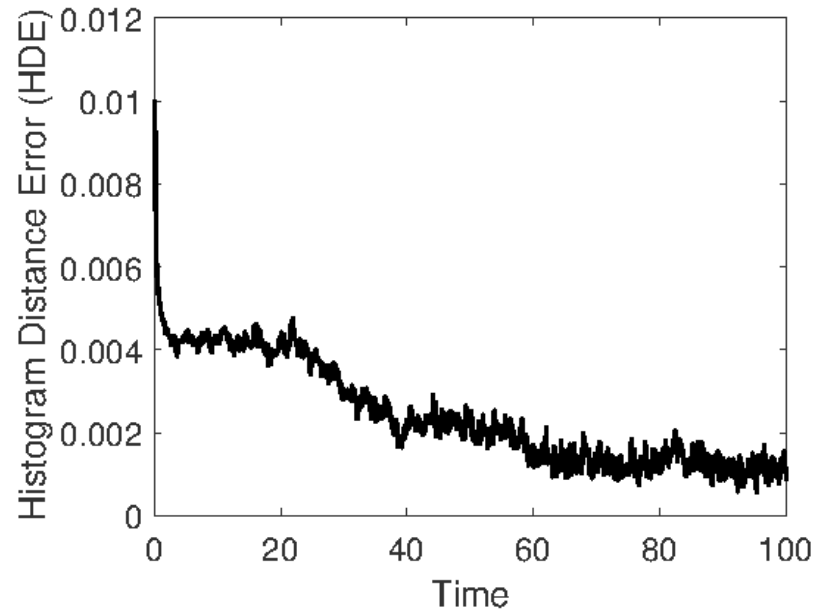
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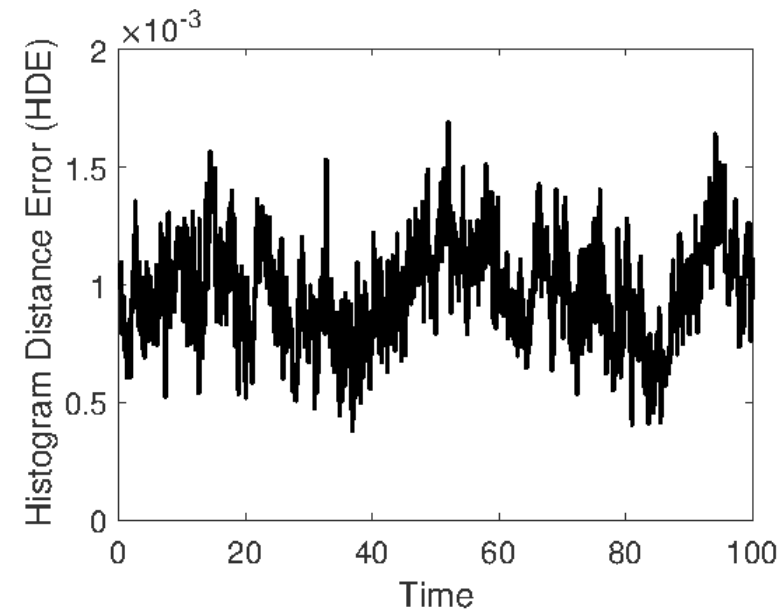


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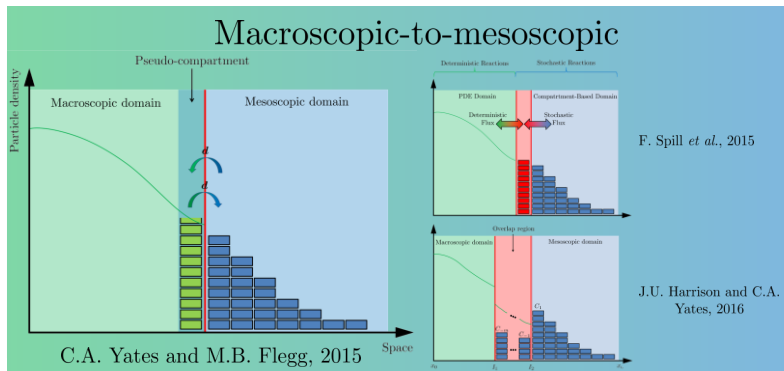
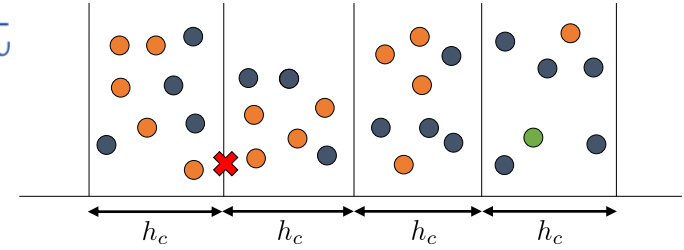
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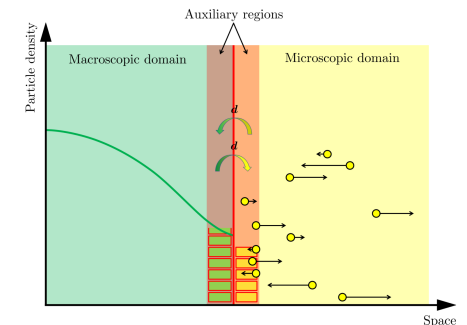
Summary

Reaction-diffusion systems may be modelled in different ways, each with complimentary advantages and disadvantages.



Hybrid methods combine these to form accurate and efficient methods. There are many examples.

The auxiliary region method combines PDE and Brownian-based approaches. Able to simulate reaction-diffusion systems accurately.



Thank you for your attention.


Any questions?

This work is joint with Kit Yates.



Get in touch:

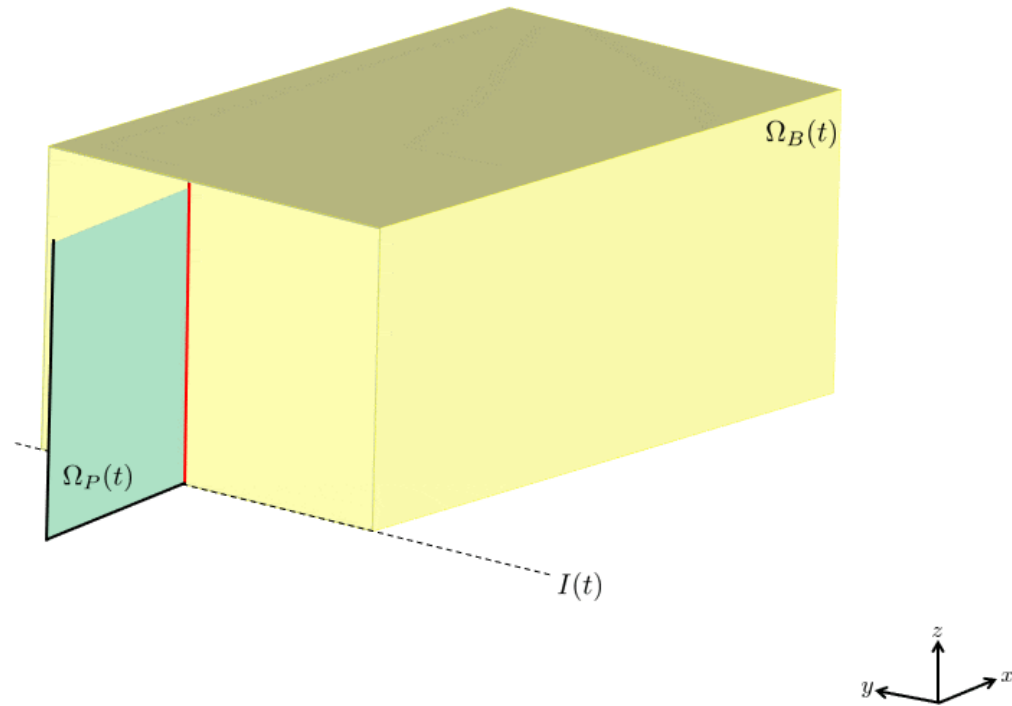
 @C_A_Smith50

 c.smith3@bath.ac.uk

 <https://people.bath.ac.uk/cs640/>

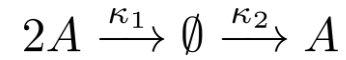
1D-3D ARM

1D-3D auxiliary region method



Results: 1D-3D ARM

Reaction system:

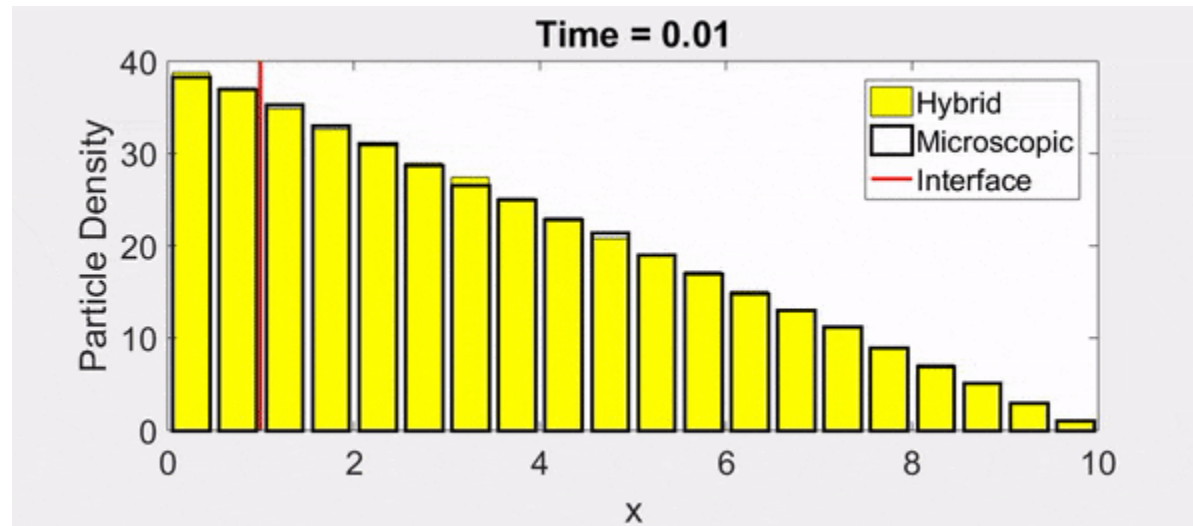


Corresponding PDE:

$$\frac{\partial u}{\partial t} = D \nabla^2 u - \frac{\kappa_1}{L_y L_z} u^2 + \kappa_2 L_y L_z$$

Moment closure (Poisson)

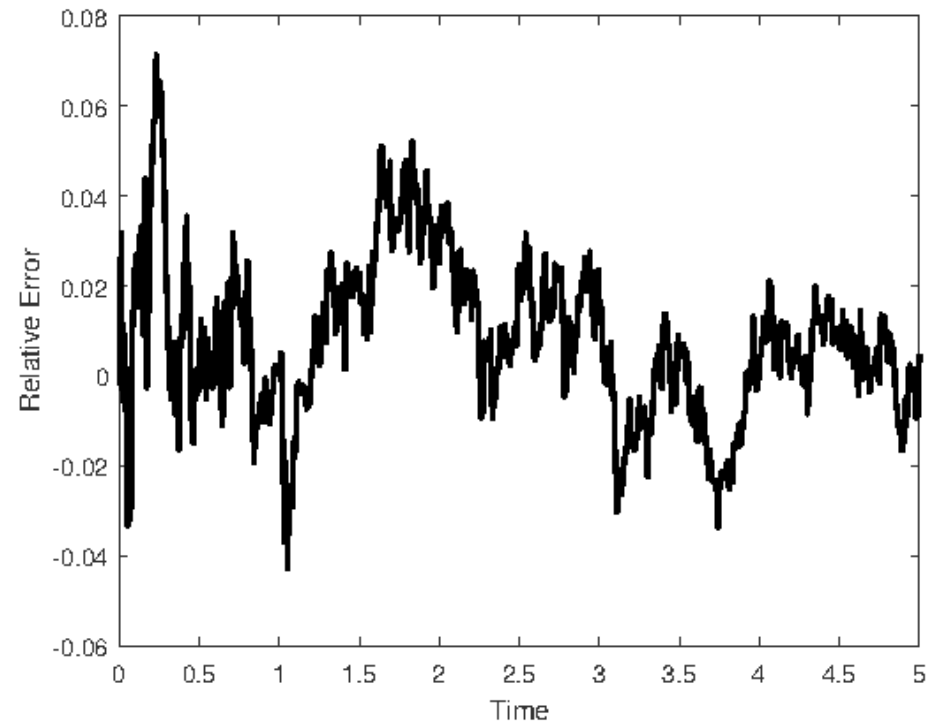
$$\langle A \rangle = \text{Var}(A)$$



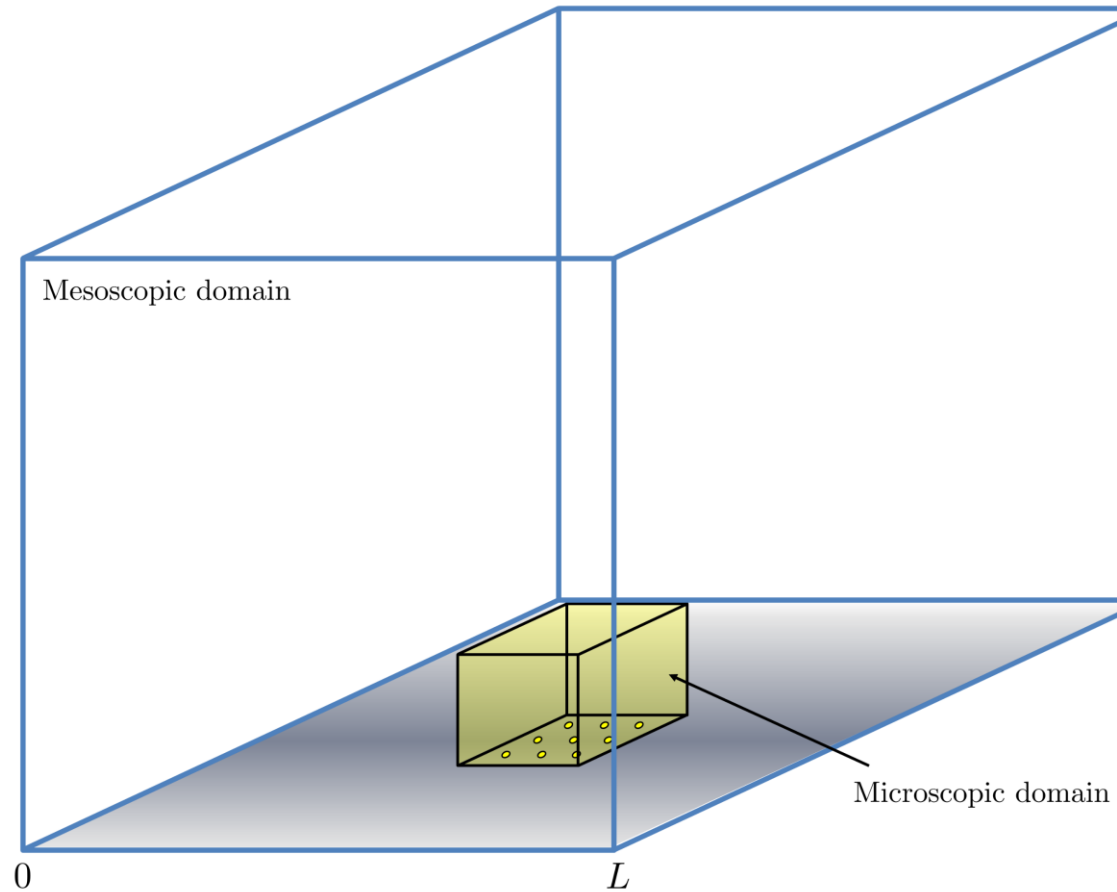
Error: 1D-3D ARM

$$E_{\text{Rel}}(t) = \frac{N_{\text{M}}(t) - N_{\text{H}}(t)}{N_{\text{M}}(t)}.$$

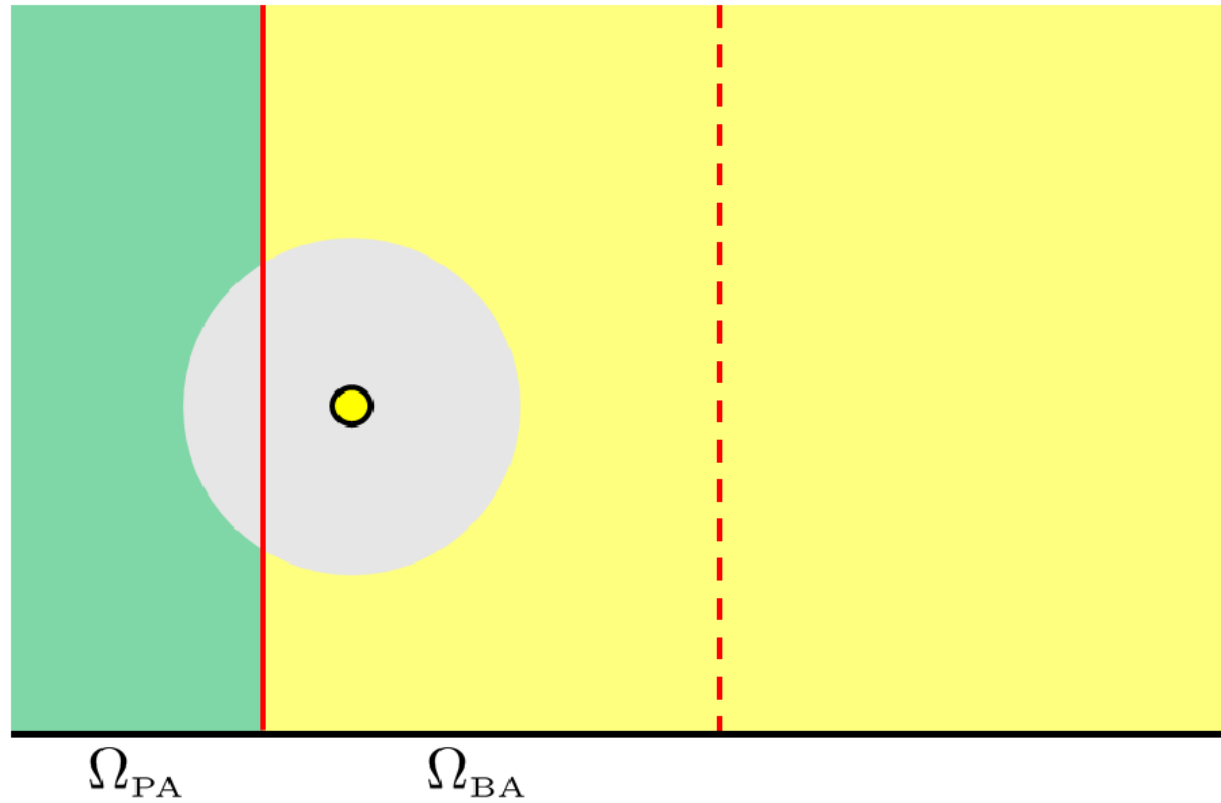
$N_{\text{H}}(t)$ is the average number of particles in the final binning width of the hybrid method at time t , and $N_{\text{M}}(t)$ is the same for the fully microscopic simulation.



Calcium induced calcium release



Reactions in the Brownian AR



Summary of models

Scale	Advantages	Disadvantages
Macroscopic (Mean-field)	<ul style="list-style-type: none">Fast to compute solutions.Suitable for high copy numbers.Amenable to analysis.	<ul style="list-style-type: none">Inaccurate for low particle numbers.Mean-field dynamics diverge from individual-level behaviour for high-order reactions.
Mesoscopic (Compartments)	<ul style="list-style-type: none">Fast for low copy numbers.Represents the individual-level behaviour.	<ul style="list-style-type: none">Can be slow for large copy numbers.Does not retain precise locations of particles or particle identity.
Microscopic (Brownian-based)	<ul style="list-style-type: none">Most accurate representation of the three.Can be used for low copy numbers.	<ul style="list-style-type: none">Slow to compute reactions.Impractical for large numbers of particles.