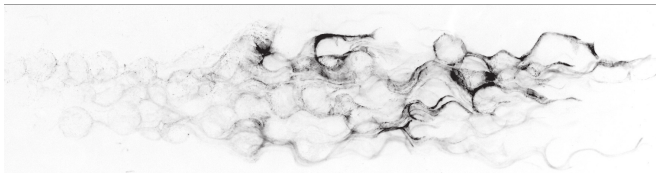


# Diffusion in arrays of obstacles: beyond homogenisation

**Alexandra Tzella**

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with

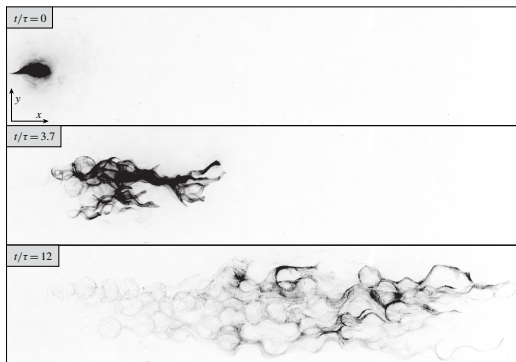
Yahya Farah, Daniel Loghin (Birmingham) and Jacques Vanneste  
(Edinburgh)

# Dispersion and mixing in media with obstacles

A multiscale problem with many science and engineering applications

- ▶ contaminant transport in soils and aquifers
- ▶ drug delivery and nutrient transport in biological tissues
- ▶ filtration devices ...

Challenge due to **advection**+**diffusion**+**geometry**.



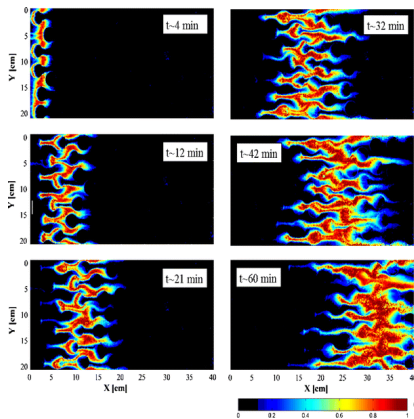
SOUZY ET AL. 2020

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## Dispersion and mixing in media with obstacles

For  $t \gg 1$ , a Gaussian approximation is used to describe how a blob evolves:

$$\theta(\mathbf{x}, t) \approx \frac{1}{\sqrt{2\pi}\mathbf{K}_{\text{eff}}t} \exp\left(-\frac{1}{4t}(\mathbf{x} - \boldsymbol{\xi}_{\text{eff}}t)^T \mathbf{K}_{\text{eff}}^{-1}(\mathbf{x} - \boldsymbol{\xi}_{\text{eff}}t)\right)$$

where  $\theta(\mathbf{x}, t)$  is the concentration and  $\boldsymbol{\xi}_{\text{eff}}$  and  $\mathbf{K}_{\text{eff}}$  are the mean velocity and the effective diffusivity tensor.

MAXWELL 1873, RAYLEIGH 1892, TAYLOR 1953, ARIS 1956, BRENNER 1981

- ▶ captures dispersion for  $\|\mathbf{x} - \boldsymbol{\xi}_{\text{eff}}t\| = O(t^{1/2})$ .

PAVLIOTIS AND STUART 2007

- ▶ do better, with **large deviations** e.g. TOUCHETTE 2009:

$$\theta(\mathbf{x}, t) \asymp \exp(-tg(\mathbf{x}/t)).$$

HAYNES & VANNESTE 2014

# The basic problem: diffusion in the presence of obstacles

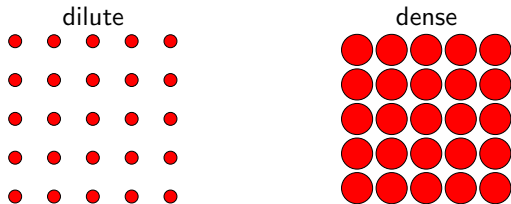


Figure: An example of a medium with circular obstacles arranged in square arrays (red).

$$\frac{\partial \theta}{\partial t} = \nabla^2 \theta,$$
$$0 = \mathbf{n} \cdot \nabla \theta, \quad \mathbf{x} \text{ on } \mathcal{B},$$

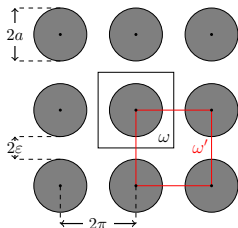
where  $\mathbf{n}$  is the unit normal to the boundaries of the obstacle  $\mathcal{B}$ .

## Macroscopic behaviour: large deviations

To capture both the Gaussian core and the tails, take the two-scale form

$$\theta(\mathbf{x}, t) \sim t^{-1} \phi(\mathbf{x}) e^{-tg(\xi)}, \quad \text{where } \xi = \frac{\mathbf{x}}{t} \in \mathbb{R}^2 \text{ and } \mathbf{x} \in \omega$$

HAYNES AND VANNESTE 2014



- ▶ **rate function**  $g$  captures dispersion for  $|\mathbf{x}| = O(t)$
- ▶ find  $g$  by solving a family of **cell eigenvalue problems**

# Cell eigenvalue problem

Leading-order problem satisfies

$$\begin{aligned}\nabla_{\mathbf{x}}^2 \phi - 2\mathbf{q} \cdot \nabla_{\mathbf{x}} \phi + |\mathbf{q}|^2 \phi &= f(\mathbf{q})\phi, \\ \mathbf{n} \cdot [\nabla_{\mathbf{x}} \phi - \phi \mathbf{q}] &= 0, \quad \mathbf{x} \text{ on } \mathcal{B} \\ \phi &\text{ periodic in } \mathbf{x}.\end{aligned}$$

where

$$\mathbf{q} = \nabla_{\xi} g(\xi) \quad \text{and} \quad f(\mathbf{q}) = \sup_{\xi} (\xi \cdot \mathbf{q} - g(\xi)).$$

- ▶ Principal eigenvalue determines the **rate function**  $g(\mathbf{x}/t)$  by taking a Legendre transform.

## Macroscopic behaviour: effective diffusion

- ▶ For  $|\mathbf{x}| \ll t$ ,

$$g(\mathbf{x}/t) \sim \frac{1}{2}(\mathbf{x}/t)^T \nabla_{\mathbf{x}/t} \nabla_{\mathbf{x}/t} g(0)(\mathbf{x}/t) = \frac{1}{4} \kappa_{\text{eff}}^{-1} (|\mathbf{x}|/t)^2, \quad (1)$$

where  $\kappa_{\text{eff}}$  is the effective diffusivity.

- ▶ Introducing (1) inside

$$\theta(\mathbf{x}, t) \asymp \exp(-tg(\mathbf{x}/t))$$

recovers Gaussian approximation for  $\theta$  obtained via homogenisation.



## Effective diffusivity in the dilute limit

In periodic arrays  $\kappa_{\text{eff}}$  was first computed using **Rayleigh's multipole method**:

$$\kappa_{\text{eff}} \sim 1 - \sigma, \quad \text{as } \sigma \rightarrow 0$$

where  $\sigma$  is the **solid area fraction**.

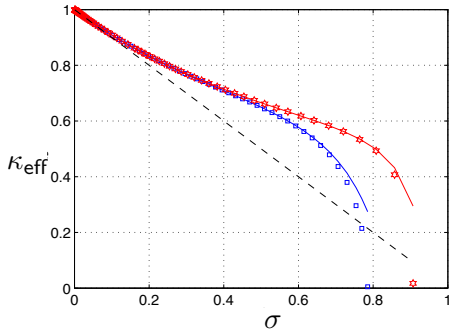


Figure: Square vs hexagonal (red) lattice.

## Effective diffusivity in the dense limit

Keller's total flux inside the narrow gaps between obstacles:

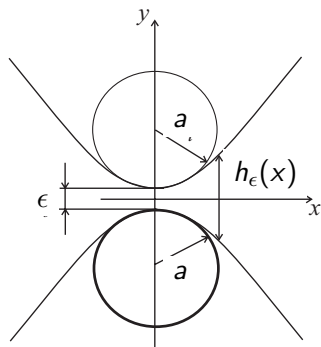
$$F = h_\epsilon(x) \frac{\partial \theta}{\partial x},$$

where the gap width

$$h_\epsilon(x) \approx \frac{1}{\pi} x^2 + \epsilon \quad \text{for } \epsilon \ll 1.$$

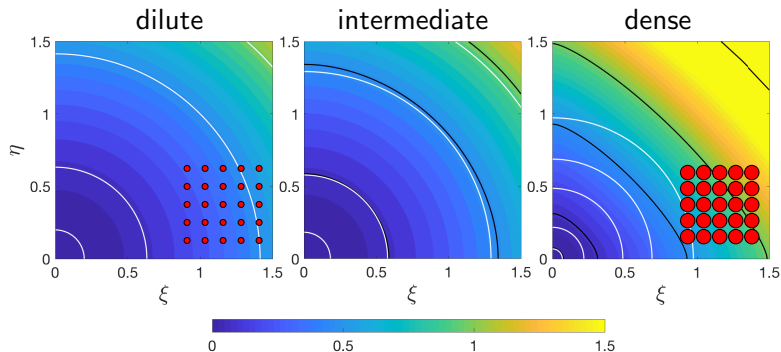
Divide by  $h_\epsilon$  and integrate:

$$F = \alpha \Delta \theta, \quad \text{where } \alpha = \sqrt{2\epsilon/\pi^3}.$$



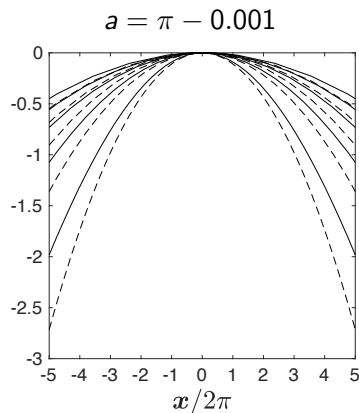
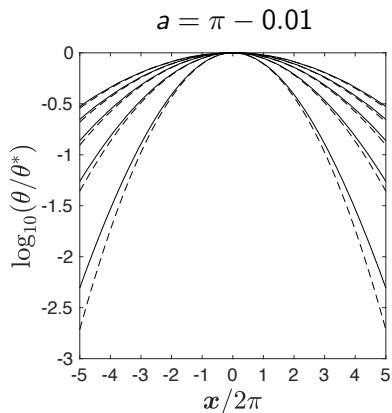
$$\kappa_{\text{eff}} \sim \frac{2\sqrt{2}\pi^2}{\mathcal{A}} \alpha = \frac{2}{\pi^{3/2}} \frac{(\pi/4 - \sigma)^{1/2}}{1 - \sigma}, \quad \text{as } \sigma \rightarrow \pi/4.$$

# Circular obstacles in square lattices: rate function $g(\mathbf{x}/t)$



- ▶ dilute case: quadratic (Gaussian) approximation (white lines) excellent for  $|\mathbf{x}| = O(t)$
- ▶ more generally: does not capture the **anisotropic** behaviour for  $|\mathbf{x}| = O(t)$

# Circular obstacles in square lattices: concentration $\theta(\mathbf{x}/t)$



- ▶ tail concentrations much fatter than predicted by the effective diffusion approximation
- ▶ discrepancy largest at earlier times and bigger radii

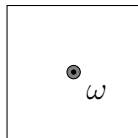
## Asymptotic analysis of the cell eigenvalue problem.

Let  $\psi = e^{-\mathbf{q}\cdot\mathbf{y}}\phi$ . The eigenvalue problem becomes the **modified Helmholtz equation**

$$\nabla_{\mathbf{y}}^2 \psi = f(\mathbf{q})\psi, \quad (2a)$$

$$0 = \frac{\partial \psi}{\partial r} \quad \text{on } r = a \quad (2b)$$

$$\psi e^{\mathbf{q}\cdot\mathbf{y}} \quad 2\pi\text{-periodic} \quad (2c)$$



### Dilute limit

- ▶ *outer region*:  $r = O(1)$ ,  $\psi \sim e^{-\mathbf{q}\cdot\mathbf{y}}$
- ▶ *inner region*:  $R = r/a = O(1)$ ,

$$\Psi \sim 1 - a|\mathbf{q}|((R + \boxed{R^{-1}})) \cos(\theta - \alpha),$$

where  $\mathbf{q} = |\mathbf{q}|(\cos \alpha, \sin \alpha)$ .

Multiply (1a) by  $e^{\mathbf{q}\cdot\mathbf{y}}$  and integrate:

$$f(\mathbf{q}) \sim \kappa_{\text{eff}}|\mathbf{q}|^2 \Rightarrow g(\mathbf{x}/t) \sim |\mathbf{x}|^2/(4\kappa_{\text{eff}}t)$$

$\Rightarrow$  effective diffusion accurate for  $|\mathbf{x}| \sim O(t)$

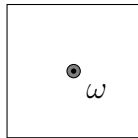
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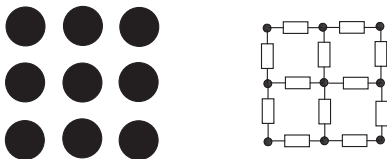
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## Dense limit: discrete network model

Use Keller's flux to build a **discrete network model**:

$$\mathcal{A} \frac{d\theta_{m,n}}{dt} = \alpha(\theta_{m+1,n} + \theta_{m,n+1} + \theta_{m-1,n} + \theta_{m,n-1} - 4\theta_{m,n}).$$



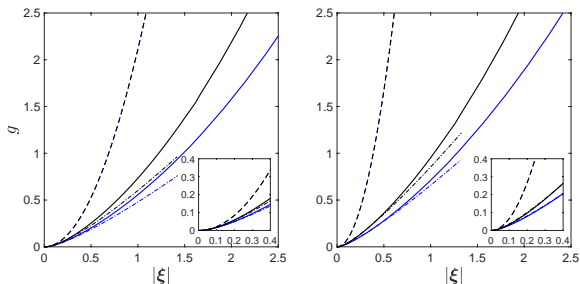
Taking  $\theta_{m,n} \sim t^{-1} \exp(-tg_d(\mathbf{r}_{m,n}/t))$  yields

$$g_d(\boldsymbol{\xi}) = \frac{2\alpha}{\mathcal{A}} (S(\beta\xi) + S(\beta\eta)),$$

where  $S(x) = 1 + x \sinh^{-1} x - \sqrt{1 + x^2}$  and  $\beta = \mathcal{A}/(4\pi\alpha)$ .

## Dense limit: discrete network model

Rate function  $g$

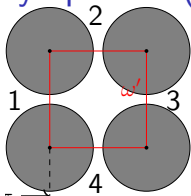


**Figure:** Rate function  $g$  against  $|\xi| = |\mathbf{x}|/t$  for (left)  $\epsilon = 0.01$  and (right)  $0.001$  in the directions  $(1, 1)$  (black) and  $(1, 0)$  (blue). Numerical results (thick solid lines), quadratic (Gaussian) approximation (dashed lines) and the network approximation (dashed-dotted).

- Captures part of the rate function and thus the tails



## Dense limit: matched asymptotics (1)



Inner gap regions:  $X = x/\sqrt{\epsilon} = O(1)$ ,  $Y = (y + \pi)/\epsilon \sim \pm H(X)$

Leading-order inner solution satisfies  $\partial_X(H(X)\Psi_0) = 0 \Rightarrow$

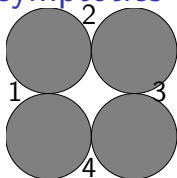
$$\Psi_0 = \alpha_1 \int_0^X \underbrace{\frac{dX}{X^2/(2\pi) + 1}}_{H(X)} + \beta_1 = \alpha_1 \tan^{-1}(X/\sqrt{2\pi}) + \beta_1,$$

and similarly for the solution in the other 3 gaps.

+ “tilted” periodicity:

$$(\alpha_3, \beta_3) = e^{-2\pi p}(\alpha_1, \beta_1) \quad \text{and} \quad (\alpha_4, \beta_4) = e^{-2\pi q}(\alpha_2, \beta_2)$$

## Dense limit: matched asymptotics (2)



**Outer region:**  $x = O(1)$ ,  $y = \pm h(x)$  where  $h(x) = x^2/(2\pi)$

Leading-order outer solution satisfies  $\partial_x(h(x)\psi_0) = 0 \Rightarrow$

$$\psi_0 \sim \gamma_1 x^{-1} + \delta_1 \quad \text{as } \mathbf{x} \rightarrow \mathbf{x}_1$$

and similarly for the other 3 gaps.

Canonical problem:  $\nabla^2 \psi^* = f \psi^*$ ,  $x^2 \partial_x \psi^* \rightarrow 1$  as  $\mathbf{x} \rightarrow \mathbf{x}_1$

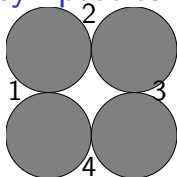
$$\psi^* \sim -x^{-1} - D_1(f) \quad \text{as } \mathbf{x} \rightarrow \mathbf{x}_1$$

$$\psi^* \sim -D_i(f) \quad \text{as } \mathbf{x} \rightarrow \mathbf{x}_i \text{ for } i = 2, 3, 4$$

$$\psi_0 = \gamma_1 \psi^* + \gamma_2 \mathcal{R}_{\pi/2} \psi^* + \gamma_3 \mathcal{R}_{\pi} \psi^* + \gamma_4 \mathcal{R}_{3\pi/2} \psi^*$$

**Matching:** Linear system, 12 unknowns  $\Rightarrow$  trans. equation  $\Rightarrow f(\mathbf{q})$

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## Dense limit: matched asymptotics

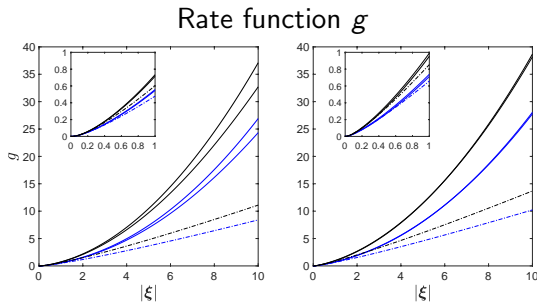


Figure: Numerical results (thick solid lines), matched asymptotic approximation (thin solid lines) and discrete-network approximation (dashed-dotted).

- ▶ discrete network approximation as a limit of the matched asymptotic prediction
- ▶ analysis breaks down for  $|\mathbf{x}| \gg t$  when

$$\theta \propto \exp(-d^2(\mathbf{x})/4t)$$

where  $d(\mathbf{x})$  is the **distance along the shortest path**  $\approx L^1$ .

## Conclusions

- ▶ large deviations generalise scalar dispersion in arrays of obstacles.
- ▶ effective diffusion underestimates concentrations at large distances/short times from the point/time of release
- ▶ effect is strongest in the dense limit, when obstacles are nearly touching.
- ▶ explicit results capture anisotropic shape of the scalar patch.
- ▶ results relevant for chemical reactions e.g. **FKPP**.

$$\frac{\partial \theta}{\partial t} = \nabla^2 \theta + \alpha \theta (1 - \theta),$$
$$0 = \mathbf{n} \cdot \nabla \theta, \quad \mathbf{x} \text{ on } \mathcal{B},$$

The front speed is deduced from  $\boxed{g(c(\mathbf{e})\mathbf{e}) = \alpha}$ .

FARAH, LOGHIN, TZELLA & VANNESTE, Proc. Royal Soc., 2020

Thank you for your attention!

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