

Incorporating interface permeability into the Matrix Formalism representation of the diffusion MRI signal

Zheyi Yang¹, Chengran Fang^{1,2}, Jing-Rebecca Li¹

¹Idefix, Inria Saclay, ENSTA Paris, Institut Polytechnique de Paris

²Mind, INRIA Saclay

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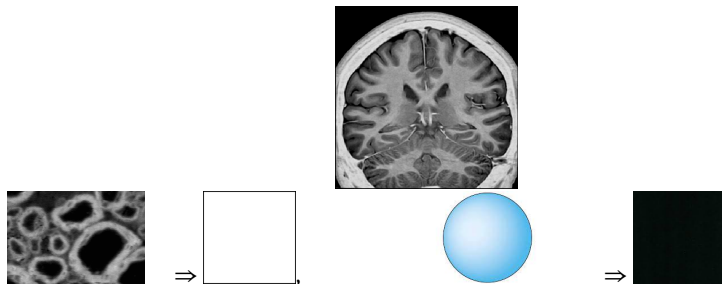


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Why permeability matters?



Traditional estimation model: Simple geometry + no water exchange between compartment \Rightarrow Inaccurate.

E.x. Estimated volume fraction will be bigger than realistic value.



Khateri M., et al. (2022) *NMR Biomed*

What does FEXI measure?

Objective

- 1 Generate fast diffusion MRI signal with different permeability values.
- 2 Propose a new reduced model, which helps permeability inference.

Bloch-Torrey equation

Bloch-Torrey equation

$$\frac{\partial}{\partial t} M^i(\mathbf{x}, t) = \nabla \cdot (D^i \nabla M^i(\mathbf{x}, t)) - I \gamma f(t) \mathbf{g} \cdot \mathbf{x} M^i(\mathbf{x}, t), \mathbf{x} \in \Omega_i, \quad (1)$$

$$D^i \nabla M^i(\mathbf{x}, t) \cdot \mathbf{n}^i(\mathbf{x}) = -D^j \nabla M^j(\mathbf{x}, t) \cdot \mathbf{n}^j(\mathbf{x}), \quad \mathbf{x} \in \Gamma_{ij}, \quad (2)$$

$$D^i \nabla M^i(\mathbf{x}, t) \cdot \mathbf{n}^i(\mathbf{x}) = \kappa^{ij} (M^j(\mathbf{x}, t) - M^i(\mathbf{x}, t)), \quad \mathbf{x} \in \Gamma_{ij}, \quad (3)$$

$$M^i(\mathbf{x}, 0) = \rho, \quad \mathbf{x} \in \Omega_i, \quad (4)$$

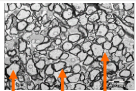
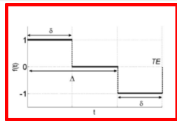
$$D^i \nabla M^i(\mathbf{x}, t) \cdot \mathbf{n}^i(\mathbf{x}) = 0, \quad \mathbf{x} \in \partial \Sigma_i. \quad (5)$$

$M^i(\mathbf{x}, t)$: Restriction of the complex transverse magnetization in Ω_i .

Conservation of water: $\kappa^{ij} = \kappa^{ji}$.

$$f(t) = \begin{cases} 1, & 0 \leq t \leq \delta, \\ -1, & \Delta \leq t \leq \Delta + \delta, \\ 0, & \text{otherwise,} \end{cases}$$

Pulsed-gradient spin echo (PGSE)



$\Omega^i, D^i \kappa^{ie} \quad \Omega^e, D^e$

N_{cmpt} compartments: $N_{cmpt} - 1$ axons and one Extracellular space (ECS);

D : Diffusion coefficient;
 ρ : initial spin.

dMRI signal

$$S(\delta, \Delta, \mathbf{g}; \kappa) = \int_{\mathbf{x} \in \bigcup \Omega_i} M(\mathbf{x}, T_E) d\mathbf{x} \quad (6)$$

- $\delta \nearrow, S(\delta, \Delta, \mathbf{g}) \searrow$;
- $\Delta \nearrow, S(\delta, \Delta, \mathbf{g}) \searrow$;
- $\|\mathbf{g}\| \nearrow, S(\delta, \Delta, \mathbf{g}) \searrow$;
- $\kappa \nearrow, S(\delta, \Delta, \mathbf{g}) \searrow$;
- Micro-structure estimation: Relate sequences parameters (δ, Δ and \mathbf{g}) to morphological parameters of interest.

A complete MRI experiment contains multiples sequences with different diffusion time, gradient strength and gradient directions.

Matrix Formalism representation

Idea: Decompose magnetization into a basis.

Theorem (Completeness of bounded Laplace operator)

Laplace operator in a bounded domain with Dirichlet, Neumann or Robin boundary condition has a complete set of eigenfunctions.

Laplace eigenmode

$$-\nabla \cdot D_i \nabla \phi_k^i(\mathbf{x}) = \lambda_k \phi_k^i(\mathbf{x}), \quad \mathbf{x} \in \Omega_i, \quad (7)$$

$$D_i \nabla \phi_k^i(\mathbf{x}) \cdot \mathbf{n}^i(\mathbf{x}) = -D_j \nabla \phi_k^j(\mathbf{x}) \cdot \mathbf{n}^j(\mathbf{x}), \quad \mathbf{x} \in \Gamma_{ij}, \quad (8)$$

$$D_i \nabla \phi_k^i(\mathbf{x}) \cdot \mathbf{n}^i(\mathbf{x}) = \kappa_{ij} (\phi_k^j(\mathbf{x}) - \phi_k^i(\mathbf{x})), \quad \mathbf{x} \in \Gamma_{ij}, \quad (9)$$

$$D_i \nabla \phi_k^i(\mathbf{x}) \cdot \mathbf{n}^i(\mathbf{x}) = 0, \quad \mathbf{x} \in \Sigma_i. \quad (10)$$

- $(\lambda_k, \phi_k(\mathbf{x}))$: k-th L^2 -normalized Laplace eigenpair on Ω ;
- Eigenbasis $\{\phi_k(\mathbf{x})\}_{k=1,2,\dots}$ orthogonal;
- $0 = \lambda_1 < \lambda_2 \leq \dots$ when $\kappa \neq 0m/s$, $0 = \lambda_1 = \dots = \lambda_{N_{cmpt}} < \lambda_{N_{cmpt}+1} \leq \dots$ when $\kappa = 0m/s$;
- $\int_{\Omega} \phi_k(\mathbf{x}) d\mathbf{x} = 0$ expect for those corresponding to 0 eigenvalue.

Matrix Formalism representation

⇒ We can decompose Bloch-Torrey operator into Laplace eigenbasis.
Rapidly oscillating eigenmodes do not contribute significantly to diffusion physics.
Length scale $l(\lambda_k) := \pi \sqrt{D_{ave}/\lambda_k}$, where $D_{ave} = \sum_{i=1}^{N_{cmpt}} D_i |\Omega_i| / |\Omega|$. Relate to wavelength.

$$M(\mathbf{x}, t) \approx \sum_{k=1}^{N_{eig}} v_k(t) \phi_k(\mathbf{x}) = \Phi(\mathbf{x}) T(t),$$

where $T(t) = [v_1(t), \dots, v_{N_{eig}}(t)]^T$, and $\Phi(\mathbf{x}) = [\phi_1(\mathbf{x}), \dots, \phi_{N_{eig}}(\mathbf{x})]$.

Define $L = \text{diag}([\lambda_1, \dots, \lambda_{N_{eig}}])$. Multiplying ϕ_k and integral over Ω ,

$$\frac{d}{dt} T(t) = -G(\mathbf{g}) T(t), \quad (11)$$

$$G(\mathbf{g}) = L + I\gamma W(\mathbf{g}),$$

$$[W(\mathbf{g})]_{kl} = \int_{\Omega} \mathbf{g} \cdot \mathbf{x} \phi_k(\mathbf{x}) \phi_l(\mathbf{x}) d\mathbf{x}, \quad (k, l) \in [[1, N_{eig}]]^2.$$

dMRI signal in Matrix Formalism representation using PGSE

$$S(\delta, \Delta, \mathbf{g}; N_{\text{eig}}, \kappa) = \int_{\Omega} M(\mathbf{x}, T_E) d\mathbf{x} = \int_{\Omega} \Phi^T(\mathbf{x}) \cdot T(T_E) d\mathbf{x} = \int_{\Omega} \Phi^T(\mathbf{x}) \cdot H(\mathbf{g}, f) \cdot T(0) \rho d\mathbf{x}, \quad (12)$$

where

$$H(\mathbf{g}, f; \kappa) = e^{-\delta \overline{G(\mathbf{g})}} \cdot e^{-(\Delta - \delta)L} \cdot e^{-\delta G(\mathbf{g})},$$

$$T(0) = \int_{\Omega} \Phi(\mathbf{x}) d\mathbf{x}.$$

Discretized form of Matrix Formalism representation

In SpinDoctor, two steps: Eigendecomposition + signal computation.
Discretized the domain with P1 basis functions $\{\varphi_p(\mathbf{x})\}_{p \in \{1, \dots, N_{nodes}\}}$.
Laplace eigenfunctions can be expressed onto P1 basis.
Multiplying $\varphi_p(\mathbf{x})$ with Laplace equation and integrating over Ω ,

Generalized matrix eigenvalue problem

Find $L \in \mathbb{R}^{N_{eig}, N_{eig}}$ and $\Phi \in \mathbb{R}^{N_{node}, N_{eig}}$ satisfying

$$(K + Q) \cdot \Phi = M \cdot \Phi \cdot L, \quad (13)$$

where M is the mass matrix, K is the stiffness matrix, Q is the interface matrix.
 $M, K, Q \in \mathbb{R}^{N_{node}, N_{node}}$.

Discretized form of Matrix Formalism representation

dMRI signal in discretized form of Matrix Formalism

$$S(\delta, \Delta, \mathbf{g}; N_{eig}, \kappa) = \left(\mathbf{1}_{N_{node},1}^T \cdot M \cdot \Phi \right) H(\mathbf{g}, f) \left(\Phi^T \cdot M \cdot \mathbf{1}_{N_{node},1} \right) \rho = T(0)^T H(\mathbf{g}, f) T(0) \rho, \quad (14)$$

where $T(0) = \Phi^T \cdot M \cdot \mathbf{1}_{N_{node},1} = [\sqrt{|\Omega_1|}, \dots, \sqrt{|\Omega_{N_{group}}|}, 0, \dots, 0]^T$. When $\kappa = 0 m/s$, $N_{group} = N_{cmpt}$.

$$H(\mathbf{g}, f; \kappa) = e^{-\delta \overline{G(\mathbf{g})}} \cdot e^{-(\Delta - \delta)L} \cdot e^{-\delta G(\mathbf{g})}.$$

$$G(\mathbf{g}) = L + I\gamma W(\mathbf{g})$$

$$[W(\mathbf{g})]_{kl} = \int_{\Omega} \mathbf{g} \cdot \mathbf{x} \phi_k(\mathbf{x}) \phi_l(\mathbf{x}) d\mathbf{x} = \phi_k^T J \phi_l, \quad (k, l) \in \llbracket 1, N_{eig} \rrbracket^2,$$

where

$$[J]_{pq} = \int_{\Omega} \mathbf{g} \cdot \mathbf{x} \varphi_p(\mathbf{x}) \varphi_q(\mathbf{x}) d\mathbf{x}, \quad (p, q) \in \llbracket 1, N_{node} \rrbracket^2.$$

- Reduce eigenvalue problem size from $N_{node} \times N_{node}$ to $N_{eig} \times N_{eig}$;
- Need to re-compute L and Φ when κ changes.


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Permeable diffusion MRI signal computation using impermeable eigenmodes

Idea: Always decompose $M(\mathbf{x}, t)$ onto the **impermeable** Laplace eigenfunctions, as unperturbed state. Consider permeability as a perturbation.

The idea similar to

 Grebenkov, Denis S. (2008) *Concepts in Magnetic Resonance*
Laplacian Eigenfunctions in NMR. I. A Numerical Tool

for permeable (Robin) boundary conditions.

The impermeable matrix eigenvalue problem:

Find $L_{imp} \in \mathbb{R}^{N_{eig}, N_{eig}}$ and $\Phi_{imp} \in \mathbb{R}^{N_{node}, N_{eig}}$ satisfying the eigenvalue matrix problem

$$K \cdot \Phi_{imp} = M \cdot \Phi_{imp} \cdot L_{imp}. \quad (15)$$

Impermeable Laplace eigenmode

$$-\nabla \cdot D_i \nabla \phi_k^i(\mathbf{x}) = \lambda_k \phi_k^i(\mathbf{x}), \quad \mathbf{x} \in \Omega_i,$$

$$D_i \nabla \phi_k^i(\mathbf{x}) \cdot \mathbf{n}^i(\mathbf{x}) = \mathbf{0}, \quad \mathbf{x} \in \Gamma_{ij},$$

$$D_i \nabla \phi_k^i(\mathbf{x}) \cdot \mathbf{n}^i(\mathbf{x}) = 0, \quad \mathbf{x} \in \Sigma_i.$$

Permeable diffusion MRI signal computation using impermeable eigenmodes

We define perturbation matrix $Q_p \in \mathbb{R}^{N_{eig}, N_{eig}}$,

$$Q_p(\boldsymbol{\kappa}) = \Phi_{imp}^T \cdot Q \cdot \Phi_{imp},$$

$$[Q_p]_{kl} = \int_{\cup_{ij} \Gamma_{ij}} \text{sgn}_{ij}(k, l) \boldsymbol{\kappa}_{ij} \phi_{imp,k}(\mathbf{x}) \phi_{imp,l}(\mathbf{x}) d\mathbf{x}, \quad (k, l) \in [[1, N_{eig}]]^2,$$

where

$$\text{sgn}_{ij}(k, l) = \begin{cases} 1, & \text{if } \phi_{imp,k} \text{ and } \phi_{imp,l} \text{ localized in the same compartment,} \\ -1, & \text{else,} \end{cases}$$

the integral over the interfaces of pair-wise impermeable Laplace eigenfunctions.

Permeable signal using impermeable eigenmodes

Theorem (dMRI signal using impermeable eigenmodes)

$$S_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = \left(\mathbf{1}_{N_{node},1}^T \cdot M \cdot \Phi_{imp} \right) H_{mod}(\mathbf{g}, f) \left(\Phi_{imp}^T \cdot M \cdot \mathbf{1}_{N_{node},1} \right) \rho, \quad (16)$$

$$H_{mod}(\mathbf{g}, f, \kappa) = e^{-\delta \overline{(L_{imp} + Q_p(\kappa) + I\gamma W_{imp}(\mathbf{g}))}} \cdot e^{-(\Delta - \delta)(L_{imp} + Q_p(\kappa))} \cdot e^{-\delta(L_{imp} + Q_p(\kappa) + I\gamma W_{imp}(\mathbf{g}))}$$

Compare to original

$$S_{per}(\delta, \Delta, \mathbf{g}; N_{eig}, \kappa) = \left(\mathbf{1}_{N_{node},1}^T \cdot M \cdot \Phi \right) H(\mathbf{g}, f) \left(\Phi^T \cdot M \cdot \mathbf{1}_{N_{node},1} \right) \rho. \quad (17)$$

$$H(\mathbf{g}, f; \kappa) = e^{-\delta \overline{(L + I\gamma W(\mathbf{g}))}} \cdot e^{-(\Delta - \delta)L} \cdot e^{-\delta(L + I\gamma W(\mathbf{g}))}.$$

Replacing original quantities with impermeable quantities

$$\Phi \rightarrow \Phi_{imp},$$

$$W(\mathbf{g}) \rightarrow W_{imp}(\mathbf{g}),$$

$$L \rightarrow (L_{imp} + Q_p), Q_p = \Phi_{imp}^T \cdot Q \cdot \Phi_{imp}.$$

Sketch of proof

(1) Using full set. $N_{eig} = N_{node}$

Define $C \equiv \Phi_{imp}^T \cdot M \cdot \Phi$. $\Rightarrow C$ is a unitary matrix, $C \cdot C^T = Id$.

Two consequences immediate:

- $\Phi_{imp} = \Phi \cdot C^T$;
- $L_{imp} + Q_p = \Phi_{imp}^T (K + Q) \Phi_{imp} = C \cdot \Phi^T (K + Q) \Phi \cdot C^T = C \cdot L_{per} \cdot C^T$.

$$\begin{aligned} S_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{node}) &= \left(\mathbf{1}_{N_{node},1}^T \cdot M \cdot \Phi_{imp} \right) H_{mod}(\mathbf{g}, f, \kappa) \left(\Phi_{imp}^T \cdot M \cdot \mathbf{1}_{N_{node},1} \right) \rho \\ &= \mathbf{1}_{N_{node},1}^T \cdot M \cdot (\Phi_{imp} \cdot C) \cdot H_{per}(\mathbf{g}, f; \kappa) \cdot \left(C^T \cdot \Phi_{imp}^T \right) \cdot M \cdot \mathbf{1}_{N_{node},1} \rho \\ &= \left(\mathbf{1}_{N_{node},1}^T \cdot M \cdot \Phi \right) H_{per}(\mathbf{g}, f; \kappa) \left(\Phi^T \cdot M \cdot \mathbf{1}_{N_{node},1} \right) \rho \\ &= S_{per}(\delta, \Delta, \mathbf{g}; N_{node}, \kappa). \end{aligned}$$

(18)

New method and Matrix Formalism are equivalent when $N_{eig} = N_{node}$.

(2) $N_{eig} < N_{node}$

$L_{imp} + Q_p = V \cdot \Lambda \cdot V^T = (C + \epsilon_2)(L_{per} + \epsilon_1)(C + \epsilon_2)^T$, $\Phi_{imp} = \Phi_{per} \cdot C^T + \epsilon_3$

$\epsilon_1, \epsilon_2, \epsilon_3 \xrightarrow{N_{eig} \rightarrow N_{node}} 0$, but not monotonously.

Properties of new method

Advantages

- Compute eigendecomposition once;
- Impermeable Laplace eigenfunctions are localized in one compartment \Rightarrow They can be computed separately.

Disadvantages

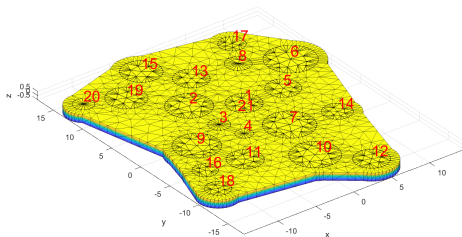
- Need more eigenmodes to reach the same error level compared to Matrix Formalism.

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Experimental Setup

- 20 axons and 1 extracellular space(ECS), connecting through permeable interface;
- volume fraction of axons is 35.0%;
- Diffusion $\sigma = 2 \cdot 10^{-3} \text{ mm}^2 / \text{s}$;
- Permeability from 10^{-5} m/s to 10^{-4} m/s , around 10^{-5} m/s for axonal membranes and myelin sheath depend on the temperature;
- 3 PGSE sequences, $\delta = \Delta$, hard case;
- Maximum g-value $< 1 \text{ T/m}$;
- Reference: Full set Matrix Formalism $S_{per}(\delta, \Delta, \mathbf{g}; N_{node}, \kappa)$;
- Criteria: Relative error $< 1\%$.



Numerical validation: Full set

Direction averaged relative error (%) =

$$100 \times 1/18 \sum_{d=1}^{18} |S_{per}(\delta, \Delta, \mathbf{g}_d; N_{node}, \kappa) - S_{new}(\delta, \Delta, \mathbf{g}_d, \kappa; N_{eig})| / S_{per}(\delta, \Delta, \mathbf{g}_d; N_{node}, \kappa),$$

where $\mathbf{g}_d = \|\mathbf{g}\| [\cos(\pi \frac{d}{18}), \sin(\pi \frac{d}{18}), 0]^T$.

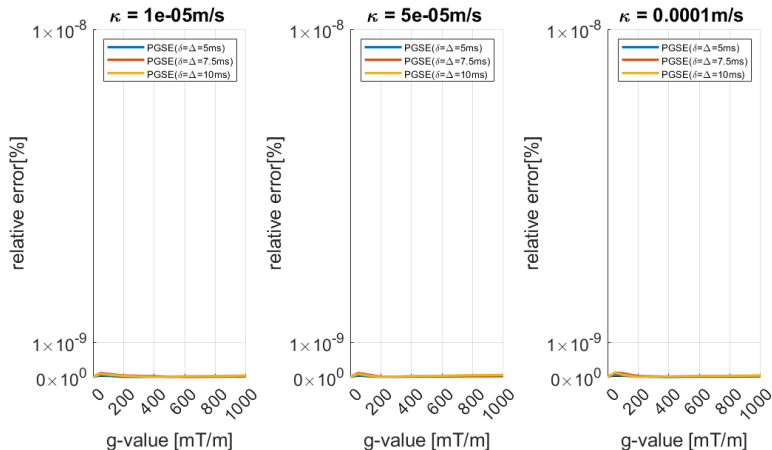


Figure: $N_{eig} = 3445$ (full set)

Numerical validation: $N_{eig} = 538$

- Proportional to diffusion time.

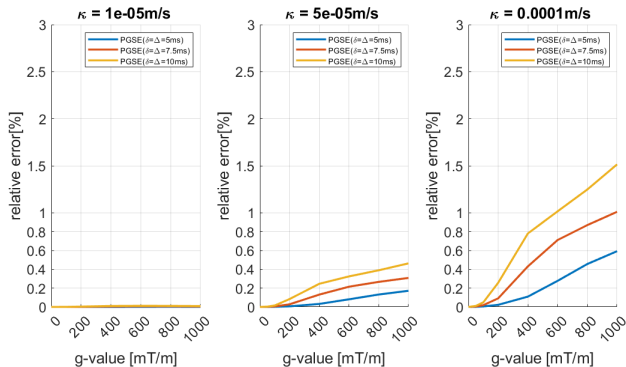


Figure: Length scale = 1 ($N_{eig} = 538$)

Numerical validation: $N_{eig} = 193$

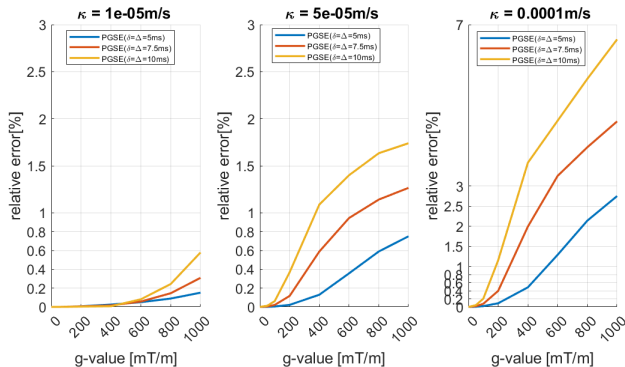


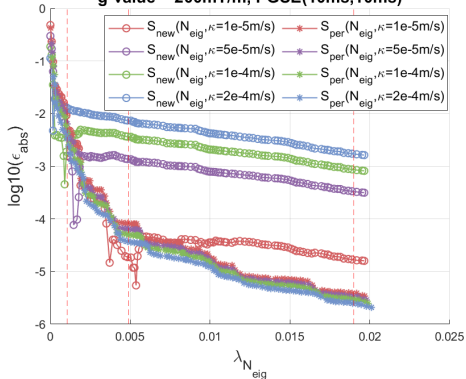
Figure: Length scale = 2 ($N_{eig} = 193$)

Convergence rate

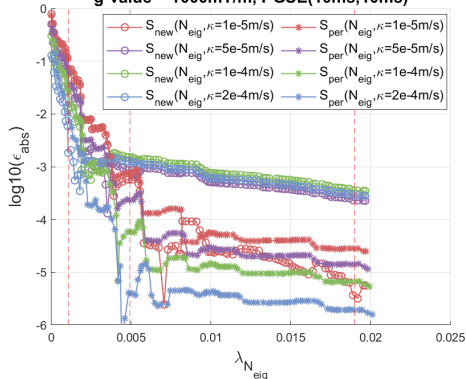
$$\epsilon_{abs} = |S_{per}(\delta, \Delta, \mathbf{g}; N_{node}, \kappa) - S_i(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})| / |\Omega|, i = \{new, per\}.$$

$$\text{Direction} [\sqrt{2}/2, \sqrt{2}/2, 0]^T.$$

g-value = 200mT/m, PGSE(10ms,10ms)



g-value = 1000mT/m, PGSE(10ms,10ms)

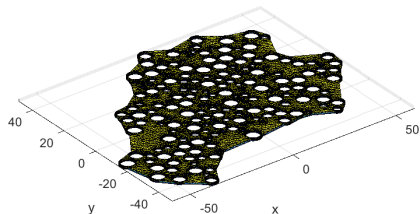
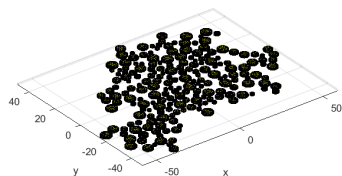


- The error decreases at the same rate for both methods in the beginning ($N_{eig} \leq 70$);
- When $N_{eig} > 70$, the error by new method decreases slower than Matrix Formalism;
- Require more eigenmodes to reach the same error level for new method.

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Computational time: eigen-decomposition



		Time(s)			
	κ (m/s)	$N_{eig} = 2000$	$N_{eig} = 4000$	$N_{eig} = 5000$	Full set
New method	0	41	171	278	141
Matrix Formalism	10^{-5}	99	419	646	723
	5×10^{-5}	100	389	518	727
	10^{-4}	102	353	625	734

Full set: 32023 nodes. ECS contains 16924 nodes.

Computational time: matrix exponential

$$\epsilon_{rel}(\%) = 100 \times \frac{|\bar{S}_{per}(\delta, \Delta, \mathbf{g}_{xy}; N_{node}, \kappa) - \bar{S}_{new}(\delta, \Delta, \mathbf{g}_{xy}, \kappa; N_{eig})|}{\bar{S}_{per}(\delta, \Delta, \mathbf{g}_{xy}; N_{node}, \kappa)},$$

where $\mathbf{g}_{xy} = \|\mathbf{g}\| [\sqrt{2}/2, \sqrt{2}/2, 0]^T$.

When N_{eig} is the same, the time is the same. \Rightarrow Save eigen-decomposition time.

κ	δ	Δ	$\ \mathbf{g}\ $	Matrix Formalism + 100s		New method + 41s		FE
				Time	ϵ_{rel}	Time	ϵ_{rel}	Time
10^{-5}	5	5	200	1.2	0.0069	1.1	0.0023	34.0
			1000	1.8	0.27	2.3	0.066	117.3
	10	10	200	1.2	0.017	1.5	0.022	53.1
			1000	3.5	0.86	3.8	0.14	199.0
10^{-4}	5	5	200	1.2	0.0063	1.0	0.11	37.7
			1000	2.7	0.21	1.7	2.6	94.6
	10	10	200	1.7	0.014	1.2	1.3	56.6
			1000	4.7	0.39	3.7	6.1	179.9

Table: $N_{eig} = 2000$. δ and Δ in μs , $\|\mathbf{g}\|$ in mT/m , κ in m/s . Direction = $[\sqrt{2}/2, \sqrt{2}/2, 0]^T$.

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Signal contribution comparison

$T_{imp}(0) = [\sqrt{|\Omega_1|}, \dots, \sqrt{|\Omega_{N_{cmpt}}|}, 0, \dots, 0]^T \Rightarrow$ Only first $N_{cmpt} \times N_{cmpt}$ items of $H_{mod}(\mathbf{g}, f, \kappa)$ are needed in signal computation.

The volume-normalized signal as

$$\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = \frac{S_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})}{|\Omega|} = \sum_{i=1}^{N_{cmpt}} \sum_{j=1}^{N_{cmpt}} \sqrt{\frac{|\Omega_i|}{|\Omega|}} [H_{mod}(\mathbf{g}, f, \kappa)]_{ij} \sqrt{\frac{|\Omega_j|}{|\Omega|}}, \quad (19)$$

where

$$H_{mod}(\mathbf{g}, f, \kappa) = e^{-\delta(L_{imp} + Q_p(\kappa) + I\gamma G_{imp}(\mathbf{g}))} e^{-(\Delta - \delta)(L_{imp} + Q_p(\kappa))} e^{-\delta(L_{imp} + Q_p(\kappa) + I\gamma G_{imp}(\mathbf{g}))}.$$

Define signal contribution as:

$$S_k(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = \sum_{i=1}^{N_{cmpt}} \sum_{j=1}^{N_{cmpt}} \left(\sqrt{\frac{|\Omega_i|}{|\Omega|}} \bar{w}_{ki} \right) e^{-(\Delta - \delta)(L_{imp} + Q_p(\kappa))} (w_{kj} \sqrt{\frac{|\Omega_j|}{|\Omega|}}), \quad (20)$$

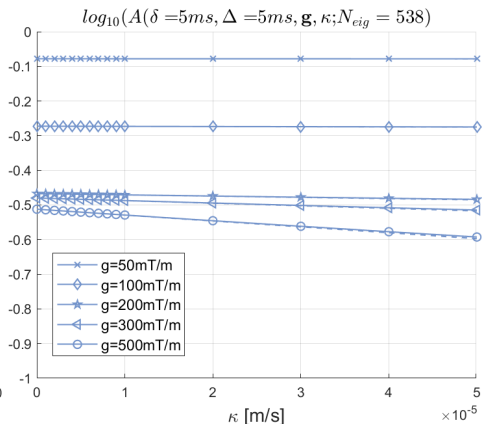
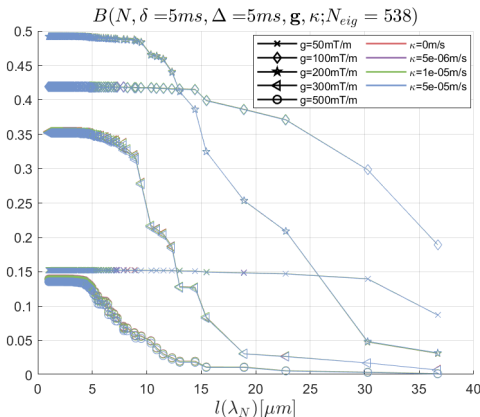
where w_{kj} is the k -th row, j -th column of the matrix $e^{-\delta(L_{imp} + Q_p(\kappa) + I\gamma G_{imp}(\mathbf{g}))}$.

$$\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = \sum_{k=1}^{N_{eig}} S_k = \sum_{k=1}^{N_{cmpt}} S_k + \sum_{k=N_{cmpt}+1}^{N_{eig}} S_k = A(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) + \sum_{k=N_{cmpt}+1}^{N_{eig}} S_k.$$

Signal contribution comparison

Define $B(N, \delta, \Delta, \mathbf{g}, \kappa; N_{eig}) := \sum_{k=N_{cmpt}+1}^N S_k, N \in [[N_{cmpt} + 1, N_{eig}]]$.

Simulated direction $[\sqrt{2}/2, \sqrt{2}/2, 0]^T$.



- $B(N, \delta, \Delta, \mathbf{g}, \kappa; N_{eig})$ is independent to κ in small permeability range;
- $A(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})$ has an exponential relation with κ in small permeability range;
- $\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = B(N_{eig}, \delta, \Delta, \mathbf{g}; N_{eig}) + A(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})$.

Approximation model for low permeable case

Reduced model for low permeability case

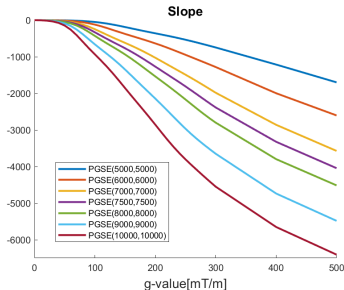
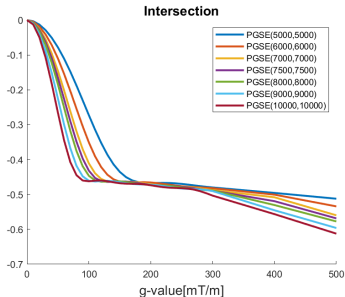
$$\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) \approx \bar{S}_{imp}(\delta, \Delta, \mathbf{g}; N_{eig}) + 10^{\alpha(\delta, \Delta, \mathbf{g})} (10^{\beta(\delta, \Delta, \mathbf{g}) \cdot \kappa} - 1) = \bar{S}_{approx}, \quad (22)$$

where $\bar{S}_{imp}(\delta, \Delta, \mathbf{g}; N_{eig})$ is the impermeable diffusion MRI signal,

$$\bar{S}_{imp}(\delta, \Delta, \mathbf{g}; N_{eig}) = B(N_{eig}, \delta, \Delta, \mathbf{g}; N_{eig}) + 10^{\alpha(\delta, \Delta, \mathbf{g})}.$$

$\alpha(\delta, \Delta, \mathbf{g})$ and $\beta(\delta, \Delta, \mathbf{g})$ are intersection and slope of $\log_{10}(A(\kappa))$, respectively.

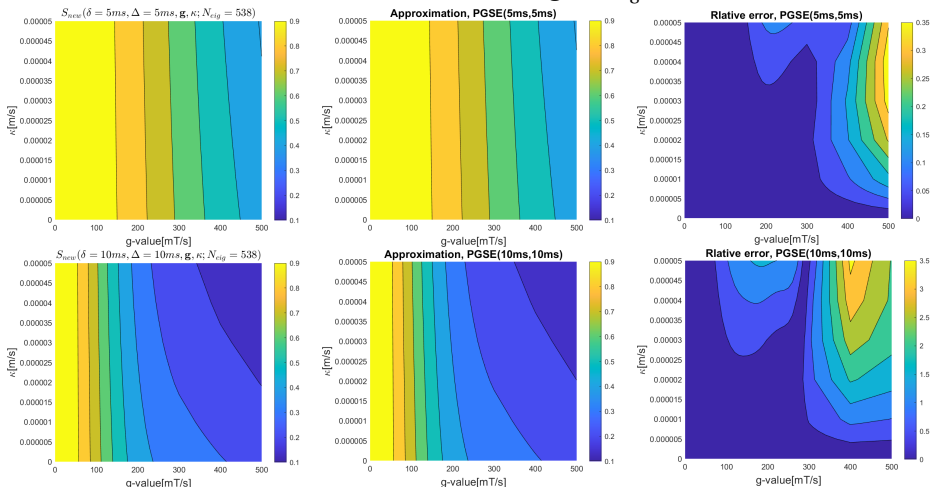
$$A(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) = 10^{\beta(\delta, \Delta, \mathbf{g})\kappa + \alpha(\delta, \Delta, \mathbf{g})}$$



Hard to get simple relation between sequence and $\alpha(\delta, \Delta, \mathbf{g})$, $\beta(\delta, \Delta, \mathbf{g})$.

Simulation

$$\epsilon_{rel}(\%) = 100 \times \frac{|\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig}) - \bar{S}_{approx}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})|}{\bar{S}_{new}(\delta, \Delta, \mathbf{g}, \kappa; N_{eig})}$$



Modified approximation model for different geometries

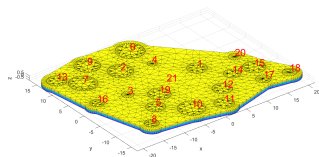


Figure: Geometry 2. Volume fraction = 25.7%.

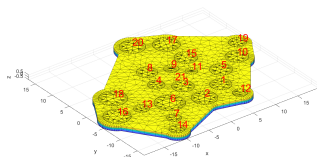


Figure: Geometry 3. Volume fraction = 32.3%.

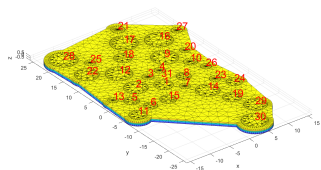


Figure: Geometry 4. Volume fraction = 31.2%.

Modified reduced model for different geometries

$$\bar{S}_{approx,geo_i} = \bar{S}_{imp,geo_i} + \frac{V_{f,geo_i}}{V_{f,geo_1}} 10^{\alpha_{geo_1}} (10^{\beta_{geo_1} \cdot \kappa} - 1), \quad (23)$$

where V_f is the volume fraction of axons. The subscript geo_i indicates the geometry i .

Simulation

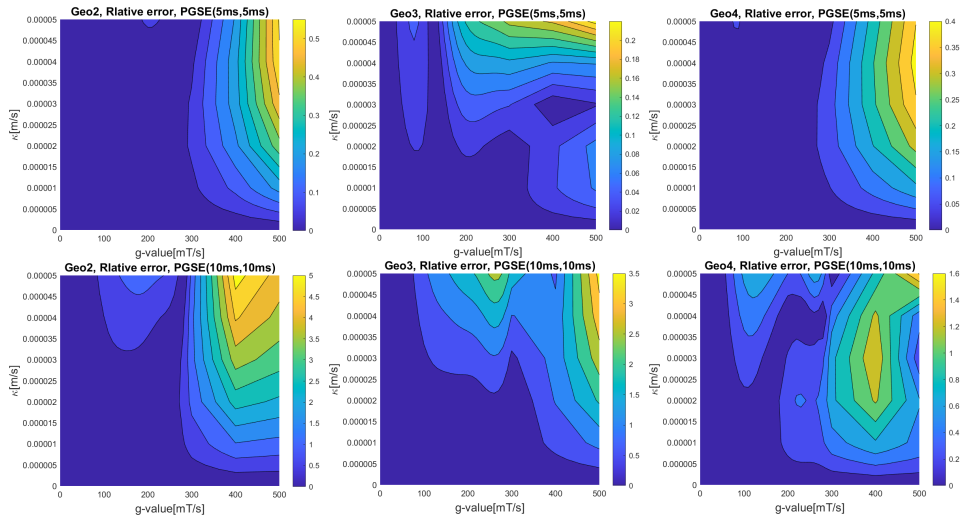


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Outlook

- A new simulation to compute diffusion MRI signal with permeable interface, using Matrix Formalism representation with impermeable Laplace eigenfunctions;
- A new reduce model yielding approximate diffusion MRI signal in the case of low permeability.

Future work

- Permeability estimation;
- Morphological parameters estimation with permeable interface.

Thank you for your attention.