

# Inferring Axon Properties with d-PGSE MRI using Analytical Water Diffusion Model

## Introduction

We present an analytical water diffusion model for inferring axon properties using low-q angular double-pulsed gradient spin echo (d-PGSE) NMR, taking into account finite gradient pulses. Estimating these properties using s-PGSE, however, requires prior knowledge of tissue orientation and high q-values, inhibiting clinical application of these methods<sup>[1]</sup>. Emerging methods for estimating orientationally invariant fibers using s-PGSE requires protocol optimization for the specific axon radius being estimated<sup>[2]</sup>. Our simulation results demonstrate that using low-q d-PGSE MRI, important axon properties including axon caliber, water diffusivity and axon volume fraction can be extracted in both orientationally known and unknown tissue.

## Method

**Imaging Protocol** The d-PGSE sequence is the simplest form of multi-PGSE, first proposed by Cory<sup>[3]</sup>. Two pairs of diffusion gradients  $G_1$  and  $G_2$  are applied at any direction with angle  $\psi$  between them. The two encoding intervals are separated by mixing time  $t_m$  with diffusion time  $\Delta_1$  and  $\Delta_2$  and, pulse duration  $\delta_1$  and  $\delta_2$ . In angular double-PGSE experiments, the diffusion time, pulse duration, and mixing time are fixed and we vary the angle  $\psi$ . Shemesh et al.<sup>[5]</sup> validated Özarslan's<sup>[4]</sup> d-PGSE signal-decay dependency theory in well controlled experiments using water-filled microcapillaries.

## Analytical Model

We propose an analytical water diffusion model for estimating axon properties based on Özarslan's theory<sup>[4]</sup> with two compartments: 1) The *intra-axonal compartment* – the space inside the axons with radius  $a$  represented by parallel non-abutting cylinders exhibits restricted diffusion, and (2) the *extra-axonal compartment* – the homogeneous substrate space outside the axons exhibits hindered diffusion. The two compartments have no water exchanged and denoted with subscript  $i$  and  $e$ , respectively. We model the **combined normalized MR signal attenuation** from the two compartments in the geometric model as:  $E = (1-f)E_e + fE_i$ , where  $f$  is axon volume fraction. We model the normalized MR signal attenuation in the extra-axonal compartment with the Gaussian diffusion distribution:  $E_e = \exp(-\gamma^2 \delta^2 D(G_1^2 + G_2^2)(\Delta - \frac{\delta}{3}))$ . The two encoding intervals have the same diffusion time and pulse

duration. We further decompose the normalized MRI signal in the intra-axonal compartment  $E_i$  into two components: parallel and perpendicular to the axon orientation:  $E_i = E_{i\parallel} \times E_{i\perp}$ . We use a discretization scheme for the gradient waveform<sup>[6]</sup> to approximate it by a train of impulses using a series of propagators and derive

$$E_{i\parallel} = \exp(-\gamma^2 \delta^2 D (G_1^2 \cos^2 \beta_1 + G_2^2 \cos^2 \beta_2)(\Delta - \frac{\delta}{3})) \quad \text{and} \quad E_{i\perp} = C + A(G_1^2 \cos^2 \beta_1 + G_2^2 \cos^2 \beta_2) + B(G_1 G_2 \cos \beta_1 \cos \beta_2), \quad \text{where}$$

- $A = 2\gamma^2 a^2 \sum_{n=1}^{\infty} S_n \times [\frac{2\delta}{\omega_n} - \frac{1}{\omega_n^2} (2 - 2e^{-\omega_n \delta} + e^{-\omega_n(\Delta-\delta)} - 2e^{-\omega_n \Delta} + e^{-\omega_n(\Delta+\delta)})]$ ;  $C = 1 - A(G_1^2 + G_2^2) - B(G_1 G_2 \cos \psi)$
- $B = 2\gamma^2 a^2 \sum_{n=1}^{\infty} \frac{S_n}{\omega_n} (e^{-\omega_n(t_m-\delta)} - 2e^{-\omega_n t_m} + e^{-\omega_n(t_m+\delta)} - 2e^{-\omega_n(\Delta+t_m-\delta)} + 4e^{-\omega_n(\Delta+t_m)} - 2e^{-\omega_n(\Delta+t_m+\delta)} + e^{-\omega_n(2\Delta+t_m-\delta)} - 2e^{-\omega_n(2\Delta+t_m)} + e^{-\omega_n(2\Delta+t_m+\delta)})$
- $S_n = \frac{1}{\alpha_n^4 - \alpha_n^2}$ ,  $\omega_n = \frac{\alpha_n^2 D}{a^2}$ ,  $\alpha_n$  are the roots of the derivatives of the first order Bessel function  $J'(\alpha_n) = 0$ .

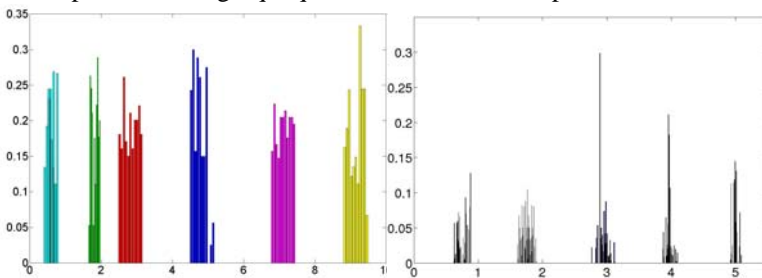
## Experiments

Our model was fitted into two diffusion experiments with Monte-Carlo simulation using Camino [7, 8]. In the first experiment, we assumed a known axon orientation aligned in the z-axis and repeated for axon radii  $a = [1, 2, 3, 5, 7, 9](\mu m)$  and  $G_{\max} = 0.5(T/m)$ . In the second experiment, we assumed an unknown orientation for axon radii  $a = [1, 2, 3, 4, 5](\mu m)$  with orientation  $\vec{u} = (1, \pi/6, \pi/3)$  and  $G_{\max} = 0.4(T/m)$ . The axon volume fraction  $f = 0.7$  and water diffusivity  $D = 2e^{-9} m^2/s$  in both experiments.

**Results** We used a Markov Chain Monte Carlo (MCMC) procedure to get samples of the posterior distribution of the model parameters given the data. Fig. 1 is our main estimation results showing the estimation-sample histograms of axon radius  $a$ . Along with axon caliber, we also extracted axon volume fraction  $f$  and water diffusivity  $D$  (data not shown)<sup>[9, 10]</sup>. Overall, we were able to extract accurate estimates of these axon properties. It is worth noticing that when axon caliber gets smaller ( $a \leq 2\mu m$ ), we observed an underestimation of the axon caliber dimension.

## Conclusions

Our estimation results demonstrate the feasibility inferring axon properties using d-PGSE without prior knowledge of tissue orientation. We conclude that modeling microstructural properties using d-PGSE acquisition may be advantageous in extracting underlying microstructural properties as it utilizes signal intensity dependency on gradient-pair direction to compensate for high-q requirement in s-PGSE experiments.



## References

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**Figure 1: Histogram of posterior distribution on axon radius a. Left: estimation in orientationally known tissue.  $a = [1, 2, 3, 5, 7, 9](\mu m)$  Right: estimation in orientationally unknown tissue.  $a = [1, 2, 3, 4, 5](\mu m)$**