

A Combined Diffusion-Perfusion Model for the Analysis of DCE-MRI Data

M. Pellerin¹, T. E. Yankeelov², and M. Lepage¹

¹Centre d'imagerie moléculaire de Sherbrooke, Université de Sherbrooke, Sherbrooke, Québec, Canada, ²Institute of Imaging Science, Vanderbilt University, Nashville, Tennessee, United States

Introduction: Many models have been proposed to describe the signal enhancement time course in tissues from images acquired with a dynamic contrast-enhanced MRI (DCE-MRI) protocol.¹⁻³ A common characteristic of the pharmacokinetic models is that they analyze the transcappillary exchange of a contrast agent (CA) on a voxel by voxel basis. In that approach, exchange of CA takes place between the blood plasma and the extravascular extracellular space. This scheme neglects the diffusion of CA within a tissue. However, CA that has extravasated in a well perfused region may diffuse to a poorly perfused, possibly necrotic, region of a tumor (Fig.1). In this case, neglecting diffusion between voxels can lead to underestimated values of the transcappillary transfer rate (K^{trans} [min^{-1}]) in the well perfused region and overestimation, even to unphysical values, of the extravascular extracellular volume fraction

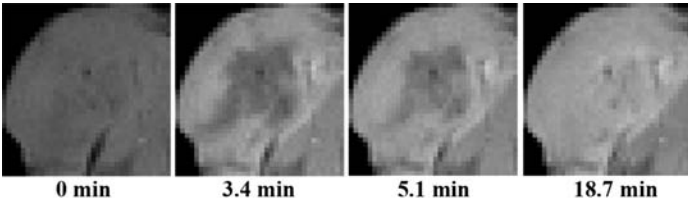


Figure 1: T1-weighted gradient echo images of mouse subcutaneous breast carcinoma. TR/TE: 200/2.4 ms, α : 30°, NA: 4, FOV: 32 x 32 mm², data matrix 128 x 128. Images have been zoomed to show the tumor only. A bolus of Gd-DTPA was injected 1 min after the first image.

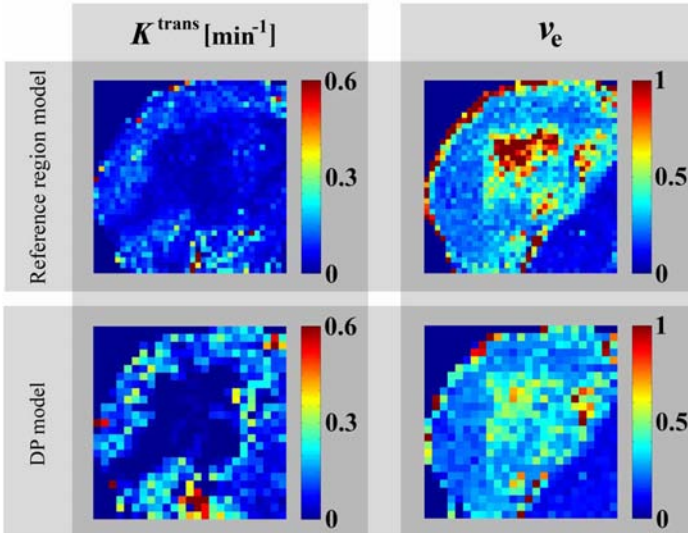


Figure 2: Fitted values of K^{trans} and v_e obtained from the reference region model and from the proposed DP model. Data used with the DP model was downsampled to reduce computation time.

was added to the data. The fitting algorithm was found to be insensitive to changes in the initial conditions. We tested our algorithm with experimental DCE-MRI data from mice acquired using the parameters stated on Fig. 1 and the apparent diffusion coefficient (ADC) of water determined from diffusion-weighted spin echo images of the same animals. As a first approximation we hypothesized that the ADC of Gd-DTPA was the same as the measured ADC of water. An arterial input function was derived from a reference region, using the formalism of ref. 3. The top row of Fig. 2 shows the values of K^{trans} and v_e as obtained from the reference region model,³ which is based on the Tofts model.¹ Unphysical values of v_e are obtained in the center of the tumor, which indicates that a standard two-compartment models may be inadequate in this case. The bottom row shows results from the DP model. Higher K^{trans} values in the well perfused periphery and lower (more realistic) values of v_e in the center are obtained. We chose to use a downsampled dataset to reduce computation time.

Conclusions: Diffusion of CA within a tissue was incorporated in a two-compartment pharmacokinetic model. This DP model was first tested using simulated data. Using real data from a mouse tumor, the DP model yielded more realistic values of v_e in the poorly perfused central region and higher values of K^{trans} in the periphery. This could increase the usefulness of K^{trans} in clinical applications.

References: 1-Tofts PS *et al.*, J Magn Reson Imaging 10 223-32 (1999) 2-Yankeelov TE *et al.*, Magn Reson Med 50 1151-69 (2003) 3-Yankeelov TE *et al.*, Magn Reson Imaging 23 519-29 (2005)

(v_e) in the necrotic region (e.g., $v_e > 1$). We propose a diffusion-perfusion (DP) model where CA diffusion is taken explicitly into account and incorporated into the standard Tofts model¹.

Methods: Adding the diffusion in 2D to the Tofts model¹ for each voxel (i,j) yields:

$$\frac{dC_{i,j}(t)}{dt} = K_{i,j}^{\text{trans}} \left(C_p(t) - \frac{C_{i,j}(t)}{v_{e,i,j}} \right) + \sum_{\text{Interface}} D(\vec{r}) \vec{\nabla} \left. \frac{C_{i,j}(t)}{v_{e,i,j}} \right|_{\text{Interface}} \frac{\vec{S}}{V}$$

where $C_p(t)$ is the plasma concentration of CA, $D(r)$ is the diffusion coefficient of the CA within the tissue, \vec{S} is the oriented surface between a voxel (i,j) of volume V and one of its neighbours. Transforming the matrix \mathbf{C} having a size of m by n into a vector \vec{C} of length $m*n$ and assuming a small time interval Δt , the solution to the differential equation is approximated by:

$$\vec{C}(t + \Delta t) = \Delta t \bar{\mathbf{K}} C_p(t) \begin{pmatrix} 1 \\ \dots \\ 1 \end{pmatrix} + \left[\mathbf{1} + \frac{1}{a^2} \Delta t \bar{\mathbf{D}} \vec{\nabla} - \Delta t \bar{\mathbf{K}} \vec{\nabla} \right] \vec{C}(t)$$

where $\bar{\mathbf{K}}$, $\bar{\mathbf{V}}$, $\bar{\mathbf{D}}$ are $m \times n$ by $m \times n$ sparse matrices, a^2 is the area of one pixel, and $\mathbf{1}$ is the $m \times n$ by $m \times n$ identity matrix.

We used a simulated annealing fitting algorithm coded in MatlabTM. A stochastic search method is needed to efficiently converge in the very large discrete solution space having countless local minima. The computation was performed on a supercomputer (872 nodes, Intel P4 with 2GB RAM per node). The performance of the proposed model was first tested on simulated data where diffusion of CA was introduced.

Results: In the simulation studies, our results show that the parameters used to generate the simulated data could be recovered reliably when realistic noise (0 – 10% of maximum concentration)