The relationship between short and long diffusion time ADC values in rat brain tumors

Alexander D Cohen¹, Peter S LaViolette², and Kathleen M Schmainda^{1,2}

¹Biophysics, Medical College of Wisconsin, Milwaukee, WI, United States, ²Radiology, Medical College of Wisconsin, Milwaukee, WI, United States

Target Audience

Brain cancer imaging researchers, radiologists and clinicians.

Purpose

Advanced imaging techniques, such as diffusion weighted imaging (DWI), have been developed to detect brain tumor progression and invasion. Biological systems are heterogeneous and contain physical boundaries that can hinder the motion of water molecules and lead to an underestimation of a material's true diffusion coefficient when the diffusion time is large. Standard clinical DWI protocols involve diffusion times (Δ) high enough for these effects to be significant (~25ms). Techniques, such as oscillating gradient spin echo (OGSE) DWI, are able to achieve diffusion times on the order of 1ms leading to more accurate diffusion measurements¹. This study looks at the relationship between short and long diffusion time apparent diffusion coefficient (ADC) values in a rat C6 glioma brain cancer model. It was hypothesized that the difference between short and long diffusion time derived ADC would be higher in tumor compared to normal tissue. This is due to increased heterogeneity and more diffusion-hindering barriers in tumor leading to restricted diffusion in the long Δ , but not the short Δ regime . These results are then compared to parameters derived from multiple b-value DWI and the stretched exponential model. This model has been shown to give estimates of intravoxel heterogeneity². Table 1. Diffusion Parameters Methods

Rats Sprague-Dawley rats were inoculated with C6 glioma cells and imaged on day 14 post-inoculation. Tumors were localized to the right hemisphere. In total, 19 rats were imaged. One rat was excluded due to poor scan quality leaving 18 rats for further analysis. Imaging All images were acquired on a 9.4T Bruker scanner. DWI ar T1-weighted images (T1+C) were acquired. For the DWI imaging, short diffusion times were obtain balanced bipolar gradients around a 180° refocusing pulse. Five sets of spin echo (SE) DWI im

Set A was acquired with Stejskal-Tanner diffusion lobes and TR/TE=2000/59.1ms, Δ=30ms, T and b-values of 0, 500, 1000, 2000, and 3000 s/mm². Sets B-E were acquired with bipolar SE DWI, TR=2000ms, b-values of 300 and 1500 s/mm², and Δ =4 or 10ms. Additional parameters are shown in Table 1. T1+C images were acquired with TR/TE=1500/7.5ms, FA=121.4°, Matrix=256x256, and Thickness=1mm. Data Analysis ADC was calculated using the standard two point model for the bipolar diffusion scans (Sets B-E). The heterogeneity index, alpha, was calculated by fitting the multiple b-value DWI data (Set A) with the stretched exponential model, $S(b)=S_0 \cdot exp(-(b \cdot DDC)^{\alpha})$, where DDC is the distributed diffusion coefficient and α is the intravoxel heterogeneity index. The intravoxel heterogeneity index ranges from 0 to 1 with lower values meaning more heterogeneity. Tumor masks were Note: ADC values are given in units of mm²/s manually drawn on the T1+C images to include areas of heightened intensity. Contralateral

normal tissue ROIs were then drawn in a similar location and size of the tumor. To examine the effects of TE and Δ on diffusion, ADC values were averaged in both the tumor and normal ROIs and compared between ROIs and scans using a repeated measures ANOVA and Tukey's multiple comparisons test. ADC-difference maps were computed by subtracting the ADC maps from Sets C and D. These maps had the same TE, but different Δ. They were then compared between ROIs using a two-sample t-test. Stretched exponential derived parameters including DDC and α, were also

averaged in the tumor and normal ROIs and compared between ROIs using a two sample t-test. Finally, correlations between ADC-difference values and a values were assessed using Pearson's correlation coefficient.

Results

Significant differences in ADC were found with both the tumor and normal cortex ROI (repeated measures ANOVA, p<0.0001 and p=0.003 respectively). Specifically, for the tumor ROI, ADC was significantly higher in sets B and C compared to sets D and E (Tukey's multiple comparisons test, p<0.01). For the normal cortex ROI, significant differences in ADC were seen between both sets B and C and set E (Tukey's multiple comparisons test, p<0.01 and p<0.05 respectively). The ADC-difference (Set B – Set C) was significantly higher in tumor vs. normal cortex (paired two sample t-test, p<0.001). The intravoxel heterogeneity index was significantly lower in tumor vs. normal cortex (paired two sample t-test, p<0.001). In addition, the ADC-difference was significantly negatively correlated with the intravoxel heterogeneity index (a) (p< 0.001). These results are summarized in Table 2 and Figure 1. Example parametric maps are shown in Figure 2.



Discussion

There were no significant differences in ADC between sequences with the same Δ , but different TE. This lends to the idea that tissue did not have compartments with different T2 relaxation times. The significant correlation between ADC-difference and a indicates ADC-difference is related to intravoxel heterogeneity. Histological studies are needed to confirm these results. These results also indicate the importance of choosing the optimal diffusion time, as optimal choice of delta may improve tumor detectability and/or tumor to brain contrast. Conclusion

Short diffusion time DWI provides a more accurate measure of ADC and allows probing smaller tissue microenvironments. The difference between short Δ ADC and long Δ ADC is related to intravoxel heterogeneity and may be useful as a tumor biomarker.

References

1. Colvin DC et al. Cancer Res. 68(14):5941-7, 2008 2. Bennett KM et al. Magn Reson Med. 50:727–734, 2003.

and contract onhono	Set B		26	4	
and contrast enhance	Set C		50	4	
images were acquired.		Set D		50	10
		Set E		70	10
Table 2. Statistical Results					
	Tumor			Normal	
ADC Difference	3.84•10 ⁻⁵		-	1.36•10 ⁻⁵	p<0.001
Alpha	0.73			0.83	p<0.001
ADC Δ =4,TE=26ms	9.33•10 ⁻⁴			7.55•10 ⁻⁴	p<0.001
ADC Δ =4,TE=50ms	9.13•10 ⁻⁴			7.38•10 ⁻⁴	p<0.001
ADC \Delta=10,TE=50ms	8.74•10 ⁻⁴			7.52•10 ⁻⁴	p<0.001
ADC Δ =10,TE=70ms	8.54•10 ⁻⁴			7.31•10 4	p<0.001

Dataset

Set A

TE (ms)

59.1

p=0.003

Δ (ms)

30