Quantitative Measurement of Microvascular Permeability in Human Brain Tumors Achieved Using Dynamic Contrast-enhanced MR Imaging: Correlation with Histologic Grade

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Quantitative Measurement of Microvascular Permeability in Human Brain Tumors Achieved Using Dynamic Contrast-enhanced MR Imaging:
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BACKGROUND AND PURPOSE: Dynamic contrast-enhanced MR imaging may be used to quantify tissue fractional blood volume (fBV) and microvascular permeability. We tested this technique in patients with brain tumors to assess whether these measurements correlate with tumor histologic grade.

METHODS: Twenty-two patients with newly diagnosed gliomas underwent MR imaging followed by surgery. Imaging consisted of one pre- and six dynamic postcontrast 3D spoiled gradient-recalled acquisition in the steady state data sets after administration of a single dose (0.1 mmol/kg) of contrast material. Signal intensity changes in blood and tissue were kinetically analyzed using a bidirectional two-compartment model, yielding estimates of fBV (mL/cm³) and microvascular permeability (mL/100 cm³ per minute). Stained tumor specimens were scored on a four-point scale (1 = low grade, 4 = high grade).

RESULTS: Histologic examination revealed one grade 1, eight grade 2, seven grade 3, and six grade 4 tumors. fBV values ranged from 0.5% to 13.7%. Permeability values ranged from 2.4 to 18.8, with a strong correlation (r = 0.83) to tumor grade. Despite some overlap between the permeability values of specific tumors from different grades, differences in the mean were statistically significant. There was a weak correlation (r = 0.39) between estimated fBV and tumor grade, and no statistically significant difference among fBV values in any of the groups.

CONCLUSION: This relatively simple method of analysis provides quantitative estimates of fBV and microvascular permeability in human brain tumors, with the permeability being predictive of pathologic grade. The technique can be easily implemented on clinical scanners and may prove useful in the assessment of tumor biology and in therapeutic trials.
assess whether these quantitative characterizations of the tumor microvasculature correlate with histologic grade.

Methods

Patients

Twenty-two patients with gliomas were included in the study. Nine patients were male and 13 were female; ages ranged from 14 to 79 years. All patients underwent surgical biopsy or tumor removal within 7 days after the MR examination.

Histologic Examination

Tumor specimens were stained with hematoxylin-eosin and were graded qualitatively by a pathologist, who was blinded to the imaging findings, and classified into four groups: grade 1 (eg, low-grade dysembryoblastic neuroepithelial tumor); grade 2 (eg, astrocytoma, oligodendroglioma, or mixed oligoastrocytoma); grade 3 (eg, anaplastic astrocytoma or oligodendroglioma); and grade 4 (eg, high-grade glioblastoma multiforme).

MR Imaging

MR examinations were performed on a 1.5-T unit. A sagittal T1-weighted scout scan was obtained to localize the tumor and to appropriately place the dynamic series. For microvascular measurements, an axial 3D spoiled gradient-recalled acquisition in the steady state (SPGR) sequence was obtained with the following parameters: 8.3/1.5/1 (TR/TE/excitations); flip angle, 30°; matrix, 256 × 128 × 28; section thickness, 3 mm; field of view, 24 cm; and acquisition time, 26 seconds per volume of 28 slices. This 3D data set was acquired seven times, once before and six times immediately after intravenous bolus administration of a single dose of contrast material (0.1 mmol/kg gadodiamide). The contrast agent was administered either by hand injection or with a power injector (Medrad, MA) with a flow of 2 mL/s.

Data Analysis

Signal intensity values measured over time in regions of interest (ROIs) in blood and tumor tissue were analyzed using the MRVision display/analysis software (The MRVision Co, Menlo Park, CA). The blood data were usually measured from the sagittal sinus; however, if the mass effect of the tumor compressed the sinus, another relatively large vascular reference was chosen. The tumor was divided into different areas (eg, rim versus core, infiltrating tumor, etc) and ROIs were measured in those different tumor portions. ROIs were selected from the last postcontrast image: a semiautomated threshold-based technique was used to identify the pixels in the tumor that had the 10% to 20% highest signal intensity in that image (which resulted from a combination of high blood volume and high microvascular permeability). Thus, the ROIs were selected with an irregular contour to cover the whole of the chosen area, without the constraints of a particular geometry.

The signal intensity changes (ASI = Spost - Spre) were calculated for the different ROIs for each postcontrast time point, and their time course was used for subsequent kinetic analysis. Signal intensity changes were recognized only as an approximation of gadolinium tracer concentration, as has been observed previously (13–15). Nonetheless, by use of a relatively low dose and a relatively short TE (14) and by maximizing sensitivity to small Δ1/T1 changes (16), approximate linearity was maintained.

The kinetic analysis used was a bidirectional two-compartment model based on the equations of Tofts et al (17) and others (eg, [18]), which yields estimates of fractional tissue blood volume (BV) (mL/cm3) and microvascular permeability, expressed as the transendothelial transfer constant kPS (mL/100 cm2 per minute).

In detail, the blood decay of the signal intensity was fitted to a biexponential function

\[ S_{\text{blood}} = A_1 \cdot e^{-k_1t} + A_2 \cdot e^{-k_2t} \]

yielding the parameters \(A_1\), \(A_2\), \(b_1\), and \(b_2\). Those four parameters were inserted in the following equation:

\[ S_{\text{tissue}} = fBV \cdot \left[ A_1 \frac{k_1}{(b_1-k_1)} \cdot e^{-k_1t} + A_2 \frac{k_2}{(b_2-k_2)} \cdot e^{-k_2t} \right] \]

where \(k_1\) reflects the efflux rate (permeability) from plasma to interstitium and \(k_2\) reflects the reflux rate from interstitium back to plasma.

Nonlinear regression analysis (Kaleidagraph, Abelbeck Software) yielded fit-value estimates of fBV, \(k_1\), and \(k_2\). Microvascular permeability, \(kPS\), was computed from \(k_1\), corrected
for hematocrit, and appropriately scaled; also computed was the sum of $k_1$ and $k_2$, or total permeability. For subsequent correlation with the histologic tumor grade, in cases of inhomogeneous tumor tissue, the area with the highest microvascular permeability was chosen to represent the tumor.

To create spatial maps of estimates of fBV and permeability, a simplified model was used for the automated pixel-by-pixel analysis. Blood was fitted to a monoexponential function and the kinetic model was a unidirectional model (ignoring the reflux coefficient, $k_2$) based on the equations of other investigators (18–21), which are special-case simplifications of Equation 2.

Statistical analysis of fBV and $k^{PS}$ was performed using ANOVA with post-hoc $t$-tests between histologically defined tumor groups. Furthermore, the Spearman correlation coefficient was determined between fBV and histologic grade, $k^{PS}$ and histologic grade, and fBV and $k^{PS}$. Total permeability, $k = k_1 + k_2$, was additionally evaluated in place of $k^{PS}$.

Results

The histologic evaluation of the brain tumors revealed a single grade 1 tumor (dysembryoplastic neuroepithelial tumor), eight grade 2 tumors (two astrocytomas, four oligodendrogliomas, and two mixed oligo-astrocytomas, Fig 2A), seven grade 3 tumors (six anaplastic astrocytomas and one oligodendroglioma), and six grade 4 tumors (glioblastoma multiforme, Fig 2B). Typical enhancement data from a low-grade (grade 2) and a high-grade (grade 4) tumor as well as the corresponding blood signals along with fitted curves from the kinetic model are displayed in Figure 3.

The estimated fBV values ranged from 0.5% to 13.7% (Table 1). There was a tendency toward higher blood volumes in tumors of a higher grade, but the correlation was weak ($r = .39$, Fig 4A) and there was no statistically significant difference between any of the groups (Table 2).

The $k^{PS}$ values ranged from $-0.4$ to $18.8$ (Table 1). We found a strong correlation ($r = .76$) between microvascular permeability $k^{PS}$ and tumor grade (Fig 4B). Despite some overlap between the different grades (most pronounced between grades 2 and 3), there was a statistically significant difference between $k^{PS}$ values of specific tumors from different tumor grades (3 versus 4, $P = .034$; 2 versus 4, $P = .003$; Table 2). The difference between grade 2 and grade 3 tumors was not significant ($P = .057$). Total permeability, $k$, largely paralleled $k^{PS}$, with a correlation of .83 between $k$ and tumor grade (Fig 4C). However, differences also achieved greater levels of significance between grade 2 and grade 3 tumors (2 versus 3, $P = .025$; 3 versus 4, $P = .006$; 2 versus 4, $P = .001$; Table 2) owing to the inclusion of the reflux rate term, which was greater in the more aggressive tumors, accentuating the difference between higher and lower grade tumors.

Discussion

There are several clinical demands for a noninvasive method to characterize brain tumors. Despite its widespread clinical use, histologic tumor grade alone at times can be an unreliable predictor of patients’ outcome. Furthermore, a noninvasive technique could help guide a surgeon to the most malignant spot before a biopsy. Other potential uses for permeability measurements might be for follow-up after therapy (eg, to monitor the success of chemotherapy or to differentiate radiation necrosis or postsurgical scar tissue from recurring tumor) or to noninvasively monitor the efficacy of a new treatment for gliomas, such as antiangiogenic pharmaceuticals.

The results of our study indicate that with a dynamic T1-weighted MR sequence we can quantify both the fBV and microvascular permeability in human brain tumors. Furthermore, these quantifications correlate with the histopathologic tumor grade (Fig 4) and allow statistically significant differentiation among tumor grades (Table 2).

Several studies have been reported in which first-pass MR imaging (22–24) or nuclear medicine methods, such as single-photon emission CT (SPECT) or positron emission tomography (PET) (25), were used to investigate the correlation between relative cerebral blood volume (rCBV) and tumor grading in gliomas. $^{18}$F-fluorodeoxyglucose (FDG)-PET and $^{201}$Tl-SPECT are considered to have a good specificity for tumor malignancy (26–30), but they lack a sufficient spatial resolution (resulting in insensitivity to small lesions) and suffer from insensitivity to low-grade intraaxial brain tumors (27). Furthermore, PET still lacks sufficient availability to be considered a viable clinical approach for routine application.

Contrast-enhanced MR imaging has spawned two distinct methods, based on differing philosophies, for the measurement of rCBV. Most commonly used are the first-pass techniques, based on the magnetic susceptibility contrast phenomenon and the resulting changes in $T2^*$ (22–24, 31). These were typically pioneered in studies of acute cerebral ischemia (32–35). $T2^*$ methods, albeit studied widely, have certain disadvantages. They require implementation on high-field magnets, preferably with the use of echo-planar or, with lower spatial or temporal resolution, gradient-echo sequences. Without a reference blood signal, only rCBV values can be obtained. Even with more sophisticated analyses, including arterial input function (36–38), CBV determinations remain approximate. The anatomic coverage is typically limited, and there is sensitivity to susceptibility artifacts near large vessels or osseous structures, which, for example, make it difficult to assess infratentorial tumors (31).

To alleviate some of these disadvantages, a method based on the relaxivity effect of contrast agents (ie, T1 measurements) was introduced (39, 40). Overall, image quality is better with the T1 technique but this method is limited to two slices and it relies on a sharp bolus injection of the contrast agent. Interestingly, in comparisons of the
T2* and the T1 methods, no statistically significant correlation was found between the rCBV values obtained in normal brain tissue (39). The results of all these studies, whether they are based on T2* or T1 effects of a contrast agent, have shown that increased malignancy is associated with increased blood volume and that, in principle, it might be possible to differentiate between low- and high-grade gliomas, although no statistically significant difference has been found between the rCBV values in the different tumor grades (39, 40).
Fig 2. B, MR image before contrast administration (top left) and six dynamic contrast-enhanced MR images of a grade 4 glioblastoma multiforme. Top center, 15 seconds; top right, 45 seconds; middle left, 75 seconds; middle center, 105 seconds; middle right, 135 seconds; and lower left, 165 seconds after contrast administration. From these seven images, spatial maps of fBV and k$_{PS}$ were determined: fBV (lower center) was estimated as 3.0% in the tumor rim and as 0.01% in the core. k$_{PS}$ values (lower right) were 13.6 mL/100 cm$^3$ per minute in the tumor (compared with 0.5 mL/100 cm$^3$ per minute in the core). Times are defined as the time after contrast that the center of k-space (mid-part of 3D acquisition) was attained.
The 3D-SPGR sequence we used provides high spatial resolution and anatomic coverage in exchange for low temporal resolution. Using this sequence, we observed a positive but weak correlation \( (r = .39) \) between fBV and tumor grade. It seems unlikely that low temporal resolution can be the source of the weak correlation between fBV and grade: if the first postcontrast image is acquired too late, there would most likely be leakage of some contrast agent from the blood pool into the interstitium before the first postcontrast imaging point. This effect would presumably be most pronounced in high-grade (and thus “leaky”) tumors. In general, this would result in an overestimation of the tissue fBV and an underestimation of the permeability; in particular, this overestimation of blood volume would most likely be more pronounced for high-grade tumors, by virtue of their higher permeability. Conversely, permeability estimates are likely to be most affected for high-grade tumors, with the error tending toward underestimation. Consequently, too low a temporal resolution would yield an artificial increase in fBV with tumor grade, which would rather tend to improve the correlation between fBV and grade. Conversely, the resultant underestimation of permeability as a consequence of low temporal resolution would degrade the correlation between permeability and grade.

Without compensation, use of any first-pass technique might result in an overestimation of the blood volume in the presence of a disrupted blood-brain barrier (BBB). Any small-molecular-weight contrast agent can be considered intravascular, and thus can be used for reliable blood volume quantifications only in the presence of an intact BBB.
A disrupted BBB leads to extravasation of the contrast agent in the extravascular space, even during the first pass, which consequently may lead to an overestimation of the blood volume by consideration of the permeability effect, as has been suggested for some first-pass rCBV approaches, but furthermore separately identifies the quantified microvascular permeability in order to gain more information about a given tumor. Our data show that the microvascular permeability has a much higher correlation with tumor grade than
do fBV estimates (Fig 4), and in fact allows statistical distinction among the tumor groups (Table 2).

Similar kinetic models have frequently been cited in the literature (15, 43, 44), both in the brain (17) and in the body periphery; eg, breast tumors (45). All yield estimates for fBV and permeability constants. There are, however, some subtle differences between the kinetic models. Additionally, some authors propose more empirical models to analyze dynamic data (46). In contrast to Tofts et al (45), whose permeability quantifications rely exclusively on $k_1$, but similar to Su et al (44), we consider independently the two transfer constants $k_1$ and $k_2$. Furthermore we computed their sum to give an estimate of, as we define it, the “total microvascular permeability.” This procedure resulted in better differentiation among tumor grades, especially between grades 2 and 3, so that the summated permeability parameter seems clinically more useful. It is also important to consider the small but nonnegligible contribution of the vascular compartment to tissue signal change.

Theoretically, the quantifications could be improved by calculating $\Delta(1/T_1)$ as an indicator of tracer concentration rather than by relying on signal intensity changes, $\Delta S_I$ (14). This, however, requires a precontrast $T_1$ determination and adds a time burden to the clinical routine. By maintaining low doses of contrast agent and using short TRs and TEs, nonlinearities in the relationship between $\Delta S_I$ and $\Delta(1/T_1)$ are ameliorated.

The synthesized parametric maps are, as described, based on a simplified kinetic model. Thus, the maps cannot really be used for regional quantifications; however, they facilitate immediate visualization and optimization of the ROI selection on the original images. They might potentially guide an automated region search and selection algorithm for more sophisticated analysis. Furthermore, the maps can be used to indicate the most malignant area in a heterogeneous tumor and may be used to guide biopsies. This may be particularly important in apparently low-grade gliomas, since approximately 80% of those may contain areas of higher grade anaplasia (47). Quantitative ROI analyses, however, allow high precision work-up of specific target regions. One can implement the full curve-fitting analysis, defined in Equations 1 and 2, on a pixel-by-pixel basis, but in our experience this is impractical and yields image maps that are potentially limited by noise.

We recognize that our method only assesses one aspect of tumor malignancy, its vascularity and neovasularity. There are, however, tumors with mitotic areas but no BBB disruption or angiogenic activity (48). These might account for the outliers, particularly in grade 3 and grade 4 tumors. Also there might be some fine structure or heterogeneity within the histologic grading system that accounts for some of the variation seen in $k^{PS}$ and fBV measures within tumors grouped into the same histologic grade.

The histologic finding of vascular proliferation, with or without endothelial proliferation, is associated with a poor clinical outcome (47). Thus, studies are underway that correlate the permeability measurements with vascular density, proliferation, time to tumor relapse, and patient’s outcome. It is anticipated that increased specificity for tumor grade will be achieved by combining noninvasive MR imaging estimates of microvascular permeability with other measures of tumor state, such as diffusion-weighted imaging findings (cellularity) or MR spectroscopy results (metabolic activity).

Conclusion

The use of dynamic contrast-enhanced MR imaging to quantify microvascular characteristics of brain tumors provides only a weak correlation between blood volume and tumor grade; however, the correlation between microvascular permeability and tumor grade is strong, and differences between tumors of different histologic grades are statistically significant.

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References

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