Early postoperative delineation of residual tumor after low-grade glioma resection by probabilistic quantification of diffusion-weighted imaging

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OBJECTIVE In WHO grade II low-grade gliomas (LGGs), early postoperative MRI (epMRI) may overestimate residual tumor on FLAIR sequences. Consequently, MRI at 3–6 months follow-up (fuMRI) is used for delineation of residual tumor. This study sought to evaluate if integration of apparent diffusion coefficient (ADC) maps permits an accurate estimation of residual tumor early on epMRI.

METHODS From a consecutive cohort, 43 cases with an initial surgery for an LGG, and complete epMRI (< 72 hours after resection) and fuMRI including ADC maps, were retrospectively identified. Residual FLAIR hyperintense tumor was manually segmented on epMRI and corresponding ADC maps were coregistered. Using an expectation maximization algorithm, residual tumor segments were probabilistically clustered into areas of residual tumor, ischemia, or normal white matter (NWM) by fitting a mixture model of superimposed Gaussian curves to the ADC histogram. Tumor volumes from epMRI, clustering, and fuMRI were statistically compared and agreement analysis was performed.

RESULTS Mean FLAIR hyperintensity suggesting residual tumor was significantly larger on epMRI compared to fuMRI (19.4 ± 16.5 ml vs 8.4 ± 10.2 ml, p < 0.0001). Probabilistic clustering of corresponding ADC histograms on epMRI identified subsegments that were interpreted as mean residual tumor (7.6 ± 10.2 ml), ischemia (8.1 ± 5.9 ml), and NWM (3.7 ± 4.9 ml). Therefore, mean tumor quantification error between epMRI and fuMRI was significantly reduced (11.0 ± 10.6 ml vs −0.8 ± 3.7 ml, p < 0.0001). Mean clustered tumor volumes on epMRI were no longer significantly different from the fuMRI reference (7.6 ± 10.2 ml vs 8.4 ± 10.2 ml, p = 0.16). Correlation (Pearson r = 0.96, p < 0.0001), concordance correlation coefficient (0.89, 95% confidence interval 0.83), and Bland-Altman analysis suggested strong agreement between both measures after clustering.

CONCLUSIONS Probabilistic segmentation of ADC maps facilitates accurate assessment of residual tumor within 72 hours after LGG resection. Multiparametric image analysis detected FLAIR signal alterations attributable to surgical trauma, which led to overestimation of residual LGG on epMRI compared to fuMRI. The prognostic value and clinical impact of this method has to be evaluated in larger case series in the future.

https://thejns.org/doi/abs/10.3171/2018.2.JNS172951

KEYWORDS low-grade glioma; extent of resection; postoperative imaging; diffusion-weighted imaging; medical image computing; oncology

In today’s surgical treatment of low- and high-grade gliomas, a maximal safe tumor resection is regarded as the desirable goal of surgery.⁷,¹²,¹³ For WHO grade II low-grade gliomas (LGGs), a growing body of evidence supports the positive prognostic impact of a maximized resection on progression-free and overall survival.¹⁹,¹⁴,³² However, the estimation of residual tumor after surgery to determine the extent of resection (EOR) remains challenging in predominantly nonenhancing tumors and has been addressed very heterogeneously in the past. Moreover, the
impact of tumor biology on patient prognosis and treatment response—conveyed by molecular markers such as isocitrate dehydrogenase (IDH) mutations or 1p/19q codeletion—has advanced critically, yet has been infrequently assessed in surgical outcome studies.\(^6,8,14,25\)

In nonenhancing gliomas, various studies suggest that residual tumor volume is prone to be overestimated on early postoperative MRI (< 72 hours postoperatively; epMRI) FLAIR sequences. This led to the practice of evaluating residual tumor on follow-up MRI (fuMRI) performed approximately 3–6 months after surgery to circumvent possible bias, which has been shown to more closely correlate with patient survival than epMRI.\(^4,5,9,10,14,23\) Image interpretation on fuMRI is anything but unequivocal, however, as delineating tumor progression from scarring, post-ischemic encephalomalacia, or postradiation injury can be difficult.\(^4,5,31\)

Aberrant FLAIR hyperintensity predominantly results from surgical trauma, which mixes with the hyperintense signal from the residual tumor on epMRI. This corresponds to areas of restricted diffusion on diffusion-weighted imaging (DWI) present in 31%–64% of postoperative MRI.\(^3,4,26,31,33\) Apparent diffusion coefficient (ADC) maps can be derived from DWI and enable one to identify ischemic areas as well as to delineate various brain tumors and normal white matter (NWM), because ADC inversely correlates with cellular density.\(^1,17,28,42\) ADC in LGG is usually higher than NWM and shows less variability compared to high-grade tumors. This allows for LGG delineation and stratification of LGG subtypes as astrocytoma or oligodendroglioma according to respective ADC characteristics.\(^16,36\)

We hypothesized that, using additional information conveyed by DWI, epMRI FLAIR hyperintensity can be corrected for the effects of surgical trauma so that an accurate estimation of residual tumor can already be achieved early after LGG resection. Using a probabilistic approach, we clustered corresponding ADC histograms of epMRI FLAIR hyperintense areas as residual tumor, NWM, or ischemia and compared the corrected tumor volumes to the results from fuMRI acquired approximately 3–6 months after surgery.

### Methods

#### Patient Cohort

Our IRB approved this retrospective study, and the requirement for patient informed consent was waived. All adult patients (age > 18 years) with newly diagnosed, histologically confirmed WHO grade II gliomas treated by microneurosurgical tumor resection at the Department of Neurosurgery, University Hospital Heidelberg, from 2004–2013 were screened (n = 83). Eligible patients were selected if preoperative MRI, epMRI acquired < 72 hours after surgery, and fuMRI at 3–6 months after surgery was available (n = 48). Required sequences were pre/postcontrast T1- and T2-weighted FLAIR imaging for all cases. DWI sequences were mandatory for epMRI. Only primary tumor resections were included in the analysis. Cases with early adjuvant treatment (radiotherapy and/or chemotherapy) starting prior to fuMRI were excluded (n = 5). Using these criteria, a total of 43 cases were included in our analysis.

Medical chart review was performed, including clinical parameters (sex, age at initial diagnosis), tumor location (side, lobe, eloquence), and treatment history (resection, radiation, chemotherapy).

#### Tumor Resections

Microneurosurgical tumor resections were performed by experienced neurosurgeons with the aid of neuronavigation (all cases) and intraoperative MRI (n = 31, 72%). The planned EOR was extracted from operating room records or retrospectively assessed by experienced neurosurgeons (M.S. and C.J.) in cases in which records were inconclusive. Only brain regions directly associated with motor or speech function were regarded as eloquent.

#### Imaging Parameters

Preoperative MRI and regular follow-up MRI were performed on 3-T and 1.5-T MRI machines (Verio, Trio, or Symphony; Siemens). Intraoperative MRI was performed on a 1.5-T MR machine (Espree, Siemens). Standard imaging protocols included acquisition of T1-weighted magnetization prepared rapid gradient echo (MPRAGE; 3D sagittal or axial), TI 900–1100 msec, TE 3–4 msec, TR 1710–2250 msec, FA 15°) before and after gadolinium-based contrast agent application (gadoteric acid [Gd-DOTA]), T2-weighted turbo spine echo (2D axial, TE 85–88 msec, TR 2740–5950 msec, section thickness 5.5 mm, spacing 5.5 mm), and T2-weighted FLAIR (2D axial, TE 85–135 msec, TR 8500–10000 msec, section thickness 5 mm, spacing 5.5 mm). DWI was performed using a single-shot, spin-echo, echo-planar sequence with the following parameters: TE 90 msec, TR 5300 msec, FA 90°, slice thickness 5 mm. Diffusion-sensitizing gradients were applied sequentially in the x, y, and z directions with b-values of 0 and 1000–1200 s/mm². Apparent diffusion coefficient (ADC) maps were automatically calculated by the manufacturer’s software.

#### Image Segmentation and Coregistration

Image analysis was performed using the Medical Imaging Interaction Toolkit (MITK), a free open-source research platform for medical image analysis (www.mitk.org).

Segmentation of LGG tumor volumes was performed according to FLAIR sequences for preoperative MRI, epMRI, and fuMRI. Intraoperative MR images were not subject to analysis in this study. Tumor segments were traced manually with a semiautomatic region-growing tool on axial sections and were manually corrected in coronal and sagittal reconstructions. Volumes were retrieved by summation of axial segments using the image statistics app in MITK (in milliliters [cm³]).

For analysis of DWI on epMRI, FLAIR sequences were coregistered to corresponding ADC maps. This was performed using a rigid registration algorithm based on mutual information with linear interpolation included in MITK. This approach has been reported to be data independent and to yield robust and plausible geometrical...
alignments in multimodal imaging environments.\textsuperscript{20,24} All registrations were checked for accuracy to verify valid coregistration.

**Histogram Analysis**

The segmentation of FLAIR hyperintense areas suggestive for residual tumor on epMRI was used as a volume of interest (VOI) for analysis of corresponding and coregistered ADC maps. For each VOI, an ADC histogram was created and a subclassification of tissue composing the histogram was attempted. We chose a probabilistic approach to cluster the histogram into 3 different tissue classes according to their (increasing) ADC grayscale value. We termed these tissue classes 1) ischemia, 2) NWM, and 3) residual tumor for this study. This methodology has been described previously by Simon et al., who developed it for separation and quantification of heterogeneous regions in gliomas according to their ADC and have extensively shown the reliability and reproducibility of this method.\textsuperscript{30,34} Subsequently, their work was implemented into the partial volume analysis app within MITK that was used for this analysis.

**Probabilistic Classification**

For the probabilistic classification approach, it was assumed that the distribution of the ADC among the expected tissue classes of ischemia and residual tumor could be separated into a superposition of two Gaussian curves. This would allow for calculation of the probability for each voxel to belong to either tissue class.

Using an expectation maximization (EM) algorithm, a mixture model was calculated using two superpositioned (superimposed) Gaussian curves to find the desired tissue classes of ischemia and residual tumor. Additionally, a partial volume class (NWM) was created that included inconclusively clustered voxels resulting from signal overlap among ischemia and residual tumor.\textsuperscript{30} The clustering approach is illustrated in Fig. 1C.

According to the ADC grayscale value, the EM algorithm produced posterior probabilities for each voxel to belong to either tissue class. Each voxel was weighted according to its posterior probability and the weighted sum of voxels was used to calculate the mean ADC for each class and to quantify the volume of each class. An example for image processing and analysis is given in Fig. 1.

**ADC Normalization**

Probabilistic histogram analysis was performed independently for each case. To compare ADC values of retrieved tissue classes within the study cohort, normalization was performed using a control VOI on the contralateral supratentorial NWM, preferably in the frontal lobe.

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**FIG. 1.** Illustration of image analysis workflow in a left-sided temporoinsular LGG. A: Preoperative FLAIR image with segmented tumor. B: Coregistration of epMRI FLAIR (red) with corresponding ADC maps (green). C: epMRI with segmentation of FLAIR hyperintensity. D: Corresponding ADC map with projected segments from FLAIR defining the VOI (red outline) for histogram analysis. E: fuMRI FLAIR. Note the possible postischemic encephalomalacia (green arrowhead) in contrast to suspected residual tumor (red arrowhead) in accordance with clustering results on epMRI. F: Probabilistic clustering of ADC histograms in VOI. Black line = original ADC histogram; blue line = sum of clustered Gaussian curves illustrating coherence to the original ADC histogram; green Gaussian curve = clustering for ischemia (low ADC); red Gaussian curve = clustering for residual tumor (high ADC); gray Gaussian curve = clustering for NWM (ambient ADC). C.1 and D.1: Clustering results projected on corresponding source sequences C and D (see above for color coding). Figure is available in color online only.
Mean ADC from the control VOI was used to create ADC ratios for each clustered tissue class for each patient (mean tissue class ADC/mean ADC of contralateral NWM).

Additionally, DWI was evaluated for signs of persistent areas of restricted diffusion on fuMRI, when available. Early postoperative tumor volumes retrieved from probabilistic clustering were statistically compared to the respective volumes according to regular epMRI and fuMRI FLAIR. Follow-up MRI was considered the reference measure for estimation of residual tumor in this study.

Statistical Analysis

Tumor volumes were analyzed using standard descriptive statistics. Limits of agreement (LOA) were calculated and Bland-Altman plots were drawn to assess agreement among different approaches to tumor quantification. Correlation and agreement were further assessed by Pearson correlation coefficients and concordance correlation coefficients (CCCs). For the estimates of the CCC, lower 1-sided 95% confidence intervals were calculated. A repeated measures ANOVA (followed by the Bonferroni multiple comparison test) and paired t-tests were used to test for differences in measured volumes. Those p values < 0.05 were considered statistically significant. The analyses were conducted using Prism software (version 5.0c, GraphPad Inc.).

Results

Demographics

After identification of cases based on histopathology (WHO grade II glioma) and completeness of postoperative imaging, a total of 43 patients (20 women and 23 men) with a median age of 39 years (range 18–58 years) were included in the analysis. This LGG cohort comprised 24 astrocytomas, 15 oligodendrogliomas, and 4 oligoastrocytomas according to conventional histopathology. Molecular diagnostics showed IDH mutations in 37 of 43 cases (86%). lp/19q codeletion was found in 4 of 4 oligodendrogliomas based on the modern integrated classification of gliomas in the study.27

The majority of tumors were located in the frontal lobe (n = 24, 56%) with an even distribution between left and right hemispheres. Fourteen tumors (33%) were regarded as eloquent in relation to motor or language function. All tumors were predominantly nonenhancing on MRI with 13 cases (30%) showing focal contrast enhancement. Gross-total resection (GTR) of the tumor was planned in 30 cases (70%). Postoperative imaging was performed on median day 2 after surgery (interquartile range [IQR] 1–3 days). Follow-up imaging was performed at a median of 111 days after surgery (IQR 96–156 days), which is consistent with standard imaging protocols after glioma resection.40

Volumetric Data

Mean preoperative tumor volume according to FLAIR was 32.8 ± 27.0 ml. FLAIR hyperintensity suggestive for residual tumor was 19.4 ± 16.5 ml on epMRI and 8.4 ± 10.2 ml on fuMRI. Mean FLAIR signal alteration was significantly larger on epMRI compared to fuMRI (p < 0.0001). This observation initially lead to our hypothesis that DWI could be suitable to define tissue subclasses within epMRI FLAIR hyperintensity in order to adjust for signs of surgical trauma and to prevent overestimation of residual tumor on epMRI.

The timing of epMRI (postoperative day 1, 2, or 3) did not affect mean epMRI volumes (p = 0.37). FLAIR volume decreased from epMRI to fuMRI in 42 of 43 cases (98%). An increase in FLAIR volume from epMRI to fuMRI suggested progression in 1 of 43 cases (2%).

Mean overall EOR according to fuMRI was 76% (range 36%–100%), and 82% (range 55%–100%) in cases in which GTR was planned. In this cohort, oligodendrogliomas were the larger tumors compared to astrocytomas concerning preoperative (p = 0.1), early postoperative (p = 0.4), and follow-up tumor volumes (p = 0.09), but without reaching statistical significance. Table 2 illustrates results from the volumetric analysis in this study.

Probabilistic Clustering

Probabilistic clustering was applied to distinguish residual tumor from ischemia and NWM on epMRI ADC maps. Mean posterior probabilities, which describe the probability/proportion for each tissue within the histogram, were 0.32 ± 0.22 for residual tumor, 0.48 ± 0.18 for ischemia, and 0.20 ± 0.15 for NWM (Table 2). This indicates that almost half of early postoperative FLAIR hyperintensity could be assigned to postoperative ischemia while approximately 30% was suspicious for residual low-grade tumor; 20% of FLAIR hyperintensity showed ADCs comparable to NWM and could not be classified as either tissue.

To express volume differences, the weighted sum of voxels for each tissue class was multiplied by the respective voxel volume. This subdivided mean initial epMRI FLAIR volume of 19.4 ± 16.5 ml into segments of mean ischemic volume of 8.1 ± 5.9 ml, mean residual tumor volume of 7.6 ± 10.2 ml, and mean NWM volume of 3.7 ± 4.9 ml (Table 2).

As a key finding, probabilistic clustering significantly reduced the quantification error for residual tumor between epMRI and fuMRI from 11.0 ± 10.6 ml (see above) to −0.8 ± 3.7 ml (p < 0.0001). After probabilistic clustering, residual tumor volume on epMRI no longer showed a significant difference compared with the fuMRI reference (7.6 ± 10.2 ml vs 8.4 ± 10.2 ml, p = 0.16; Fig. 2).

Coherence of Probabilistic Clustering

Probabilistic analysis of ADC histograms defined three tissue subclasses for each histogram in all cases in this study. After normalization of ADC values with their respective contralateral NWM, the mean ADC ratio was 0.76 ± 0.23 for ischemia, 1.11 ± 0.31 for NWM, and 1.48 ± 0.46 for residual tumor. Mean ADC ratios were significantly different for all three tissue classes (p < 0.0001) and matched well with respective tissue-specific ADC measures published in the literature.15–17,28,31,36 This depicts successful clustering into distinct areas of ischemia (low ADC) and residual tumor (high ADC) in relation to
NWM (ambient [neither high nor low] ADC) and represents a proof of concept for our probabilistic approach.

The mean sum of posterior probabilities across all cases was above 0.99 (0.999999982 ± 1.32876 -7), which depicts an excellent fit of the Gaussian mixture model into the respective predefined ADC histograms.

All ischemic areas were located around the resection cavity or in direct contact with the surgical field. No ischemic lesions distant to the resection cavity were observed. Evaluation of ADC maps on fuMRI (available in 31/43 cases, 72%) did not show any areas of persistent diffusion restriction in any case.

**Correlation and Agreement Analysis**

To evaluate the accuracy of probabilistic clustering for estimation of residual tumor on epMRI, correlation and agreement analysis was performed with fuMRI serving as the reference measure (Fig. 3). Residual tumor volumes retrieved by probabilistic clustering showed good correlation and agreement with fuMRI estimates depicted by a Pearson r value of 0.96 (p < 0.0001) and CCC of 0.89 (95% CI 0.83). The mean difference between both time points was −0.8 ml (95% LOA −8.1 to 6.5 ml; Fig. 3). This indicates a marginal underestimation of probabilistically clustered volumes compared to a perfect match (0-ml difference), which was not statistically significant (p = 0.16).
As shown in the Bland-Altman plot in Fig. 3, the amount of residual tumor volume was a driving force of deviation between both estimates analyzed. The largest tumors (>10 ml residual volume) produced the highest differences, which affected the 95% LOA margins. Agreement between clustered epMRI and fuMRI values by Pearson correlation coefficient, CCC, and Bland-Altman analysis in astrocytomas (r = 0.93, p < 0.0001; CCC = 0.92, 95% CI 0.85; mean difference = –1.0 ml, 95% LOA –6.2 to 4.2 ml) was comparable to agreement in oligodendrogliomas (r = 0.95, p < 0.0001; CCC = 0.84, 95% CI = 0.71; mean difference = –0.5 ml, 95% LOA –9.9 to 8.9 ml).

**Discussion**

Hypothesizing that DWI may already help to achieve reliable estimations of residual tumor volume early after LGG resection, we have used a probabilistic approach to cluster ADC maps on epMRI. Areas of postoperative FLAIR hyperintensity suggestive for residual tumor were segmented, and clustering of corresponding ADC histograms helped to delineate residual tumor. Initially, residual FLAIR volumes differed significantly between epMRI and fuMRI. Through probabilistic clustering, however, effects of postoperative ischemia evident on epMRI could successfully be adjusted for, leading to close approximation of residual tumor volume to the volume observed on fuMRI. Correlation and agreement analysis with fuMRI underscore the feasibility of tumor volume estimation early after LGG resection.

As a consequence of the surgical trauma, MRI signal alterations attributable to a postresection injury have been described as time-dependent contrast enhancements on T1 and expression of FLAIR hyperintensity on MRI early after surgery.\(^2,4,10,31,37\) Over the course of time, scarring and the development of posts ischemic encephalomalacia or postradiation injury render accurate delineation of residual tumor or detection of tumor progression challenging, especially in nonenhancing LGGs.\(^5,31\) Given the changing paradigms for the timing of adjuvant treatment in patients with LGG and the increasing role of molecular markers for diagnosis and prognosis in LGG, the necessity to achieve accurate estimates of residual tumor early after resection might arise in the near future to allow for comprehensive risk stratification of patients after surgery.\(^14,25,27,38,39\) Considering these current limitations and future aspects, we present a method that accounts for the postresection injury to already achieve accurate estimates of residual tumor early (within 72 hours) after resection of an LGG.

In general terms, a probabilistic segmentation approach increases objectivity in quantification of LGG compared to manual measurements, especially in areas of tissue inhomogeneity and blurred tissue margins as they are typically found in the early postoperative setting.\(^10,34\) By analyzing ADC histograms on epMRI, ischemic lesions could be readily identified and deducted from the total postoperative FLAIR volume. By means of the clustering algorithm, ischemic lesions were detected in all cases in this study and made up 48% ± 18% of all FLAIR hyperintensity on epMRI.

Compared to other series that divergently report ischemic lesions in 25%–75% of LGG resections, the incidence of ischemia was high in this study. We attribute this observation to our probabilistic approach that objectively aimed to identify any nontumor area based on the respective lower ADC. Despite the high incidence, however, mean ischemic volume was comparable to other reports on early postoperative ischemia.\(^3,10,31\)

At b-values of 1000 s/mm\(^2\), ADC ranges are reported from 1000–1500 \(\mu\)m\(^2\)/s for LGG and between 700 and 780 \(\mu\)m\(^2\)/s for NWM in previous studies. This implies that ADC in LGG is higher compared to NWM by a factor of about 1.3–2.4,\(^15,17,28,36\) ADC for ischemia and stroke is reported to be approximately 500 \(\mu\)m\(^2\)/s, which is inferior to NWM by a factor of 0.7.\(^19,31\) Accordingly, ADC of our clustered estimates for residual tumor (mean ADC ratio 1.48 ± 0.46) and ischemia (mean ADC ratio of 0.76 ± 0.23) matched well with the expected ranges for each respective tissue class we sought to identify, which reflects coherency of our clustering results. Tumor volumes retrieved by probabilistic clustering were statistically comparable to the reference measured on fuMRI, which illustrates that our approach mitigated previous limitations for early and accurate measurement of residual tumor after LGG resection (Fig. 2). The definition of NWM as a separate mixture class for inconclusively clustered voxels allowed us to take partial volume effects into account and enabled us to compensate for segmentation errors at the tumor margins, as well as for minor misalignments during coregistration of sequences.\(^30\)

Edema associated with infiltrative LGGs also appears as FLAIR hyperintensity and might also contribute to postoperative volume overestimation. However, edema has only played a minor role in LGG outcome studies because its delineation from LGG tumor or infiltrated tissue...
remains challenging, even with advanced imaging. Rather, FLAIR hyperintensity in general, without consideration of its subcompartments, is typically used to show survival associations. Nevertheless, our clustering algorithm should be capable of adequately considering edema on epMRI. Vasogenic edema is usually considered the predominant form of edema in brain tumors and shows increased diffusion corresponding to high ADC. Cytoxic edema, in contrast, has been shown to be the major form of edema in acute traumatic lesions of the brain and ischemia, which applies to the early postoperative setting. This form of edema presents with restricted diffusion and low ADC. Therefore, the vasogenic component of edema, which should remain stable during the early postoperative phase, should be incorporated into the tumor cluster in our analysis. Any formation of new cytotoxic edema resulting from the surgical trauma or postoperative ischemia is expected to be clustered as ischemia according to its different ADC characteristics. This implies that probabilistic clustering should identify and account for postoperative edema as a transient cause of FLAIR hyperintensity after LGG resection.

Ultimately, our clustering analysis of ADC histograms conclusively defined tissue subclasses according to ADC grayscale values. Bland-Altman comparisons as well as correlation statistics for clustered epMRI and fuMRI volumes suggest sufficient agreement between both measurements in order to accept our initial hypothesis that tumor estimation is already feasible early after LGG resection, with the additional information conveyed by DWI.

As these are preliminary observations designed to deliver a proof of concept, there are certain limitations to our results. First, this is an analysis of retrospective data. Case selection was performed from a consecutive surgical LGG cohort according to completeness of postoperative imaging studies. This implies the possibility of selection bias toward cases with more extensive resections, opposed to partial resections or tumor debulkings, in which postoperative imaging showing the obvious (i.e., residual tumor) was frequently incomplete and lead to exclusion from our retrospective study. Albeit with possible selection bias, our study illustrates how to facilitate the delineation of tumor margins from postresection injury for a wide spectrum of residual tumor volumes ranging from 1.4 to 65 ml on epMRI. Second, survival analysis could not reasonably be performed in this cohort due to the small sample size in combination with the heterogeneity of histopathology and molecular markers that critically influence survival. Our selection criteria based on imaging parameters implies a strong bias for survival analysis in this regard, and furthermore yielded subgroups too small to calculate multivariate models even after a median follow-up of more than 5 years. The effort to separately analyze astrocytomas and oligodendrogliomas in our study aimed to illustrate and support the general applicability of our image analysis methodology in nonenhancing LGG, independent from histopathology. Consequently, further evaluation of the prognostic value of our method will require a larger clinical series in the future.

In its current form, our clustering method should primarily be applied to circumscribed tumor remnants around the resection cavity. As residual tumor volumes increase in size, the natural intratumoral inhomogeneity of ADCs gains significance for the clustering algorithm and can lead to classification errors. This, in part, explains our observation made in agreement analysis, that tumor size was the main driving force for persisting differences between epMRI and fuMRI even after clustering (Fig. 3). To also generalize our method’s applicability to large tumor remnants, ADC histograms of the preoperative tumor would have to be implemented into the clustering algorithm to account for the respective intratumoral ADC signature.

After clustering of epMRI, we attempted to quantify the spatial agreement of clustering results with residual tumor on fuMRI to further support our data. Tissue reorganiza-
tion and parenchymal involution occurring from surgery until fuMRI induced tissue distortion too large to account for by rigid registration algorithms, however (see Fig. 1C vs 1E for illustration). This precluded further evaluation of spatial overlap of segments and limited our presentation of data to quantitative comparison of agreement.

Tumor progression occurring from epMRI to fuMRI could be a source of bias for our agreement analysis. In this series, only 1 case showed increase of FLAIR hyperintensity from epMRI to fuMRI, which did not affect the overall conclusion of this work, however.

Beyond this current analysis, probabilistic clustering of epMRI could also facilitate the detection of LGG progression by identifying certain regions at risk for focused evaluation on later follow-up imaging. In the same manner, clustering could have implications for planning of postoperative radiotherapy as it might provide additional information about residual FLAIR hyperintensity helping to focus radiation.

The results of this study highlight the need to assess residual tumor after LGG resection, according to multiple parameters. We point out that areas of high ADC around the resection cavity on epMRI more closely correlate with the amount of residual tumor than FLAIR hyperintensity alone. Our analysis involved a substantial amount of image processing, however, which limits its clinical applicability in the current form. Along with the developments of automated image processing and analysis, we seek to streamline our image analysis and minimize user interaction in order to make probabilistic clustering more suitable for daily practice. In clinical routine, multiparametric radiological profiling of brain tumors, called radiomics, will help to derive adjusted survival stratifications and clinical decision support systems.

Conclusions

Probabilistic segmentation of ADC maps facilitates accurate assessment of residual tumor within 72 hours after LGG resection. Multiparametric image analysis detected FLAIR signal alterations attributable to surgical trauma, which led to overestimation of residual LGG tumor on epMRI compared with fuMRI. The prognostic value and clinical impact of this method need to be evaluated in larger case series in the future.

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