The Initial Area Under the Curve Derived from Dynamic Contrast-Enhanced MRI Improves Prognosis Prediction in Glioblastoma with Unmethylated MGMT Promoter

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ABSTRACT

BACKGROUND AND PURPOSE: Although perfusion and permeability MR parameters have known to have prognostic value, they have reproducibility issues. Our aim was to evaluate whether the initial area under the time-to-signal intensity curve (IAUC) derived from dynamic contrast-enhanced MR imaging can improve prognosis prediction in patients with glioblastoma with known MGMT status.

MATERIALS AND METHODS: We retrospectively examined 88 patients with glioblastoma who underwent preoperative dynamic contrast-enhanced MR imaging. The means of IAUC values at 30 and 60 seconds (IAUC30mean and IAUC60mean) were extracted from enhancing tumors. The prognostic values of IAUC parameters for overall survival and progression-free survival were assessed with log-rank tests, according to the MGMT status. Multivariate overall survival and progression-free survival models before and after adding the IAUC parameters as covariates were explored by net reclassification improvement after receiver operating characteristic analysis for 1.5-year overall survival and 1-year progression-free survival and by random survival forest.

RESULTS: High IAUC parameters were associated with worse overall survival and progression-free survival in the unmethylated MGMT group, but not in the methylated group. In the unmethylated MGMT group, 1.5-year overall survival and 1-year progression-free survival prediction improved significantly after adding IAUC parameters (overall survival area under the receiver operating characteristic curve, 0.86; progression-free survival area under the receiver operating characteristic curve, 0.74–0.76) to the model with other prognostic factors (overall survival area under the receiver operating characteristic curve, 0.81; progression-free survival area under the receiver operating characteristic curve, 0.69; P < .05 for all) except in the case of IAUC60mean for 1-year progression-free survival prediction (P = .059). Random survival forest models indicated that the IAUC parameters were the second or most important predictors in the unmethylated MGMT group, except in the case of the IAUC60mean for progression-free survival.

CONCLUSIONS: IAUC can be a useful prognostic imaging biomarker in patients with glioblastoma with known MGMT status, improving prediction of glioblastoma prognosis with the unmethylated MGMT promoter status.

ABBREVIATIONS: AUC = area under the receiver operating characteristic curve; DCE = dynamic contrast-enhanced; EGFR = epidermal growth factor receptor; GBM = glioblastoma; IAUC = initial area under the time-to–signal intensity curve; IDH1 = isocitrate dehydrogenase 1; Ktrans = volume transfer constant; MGMT = O6-methylguanine-DNA methyltransferase; nCET = non-contrast-enhanced tumor; OS = overall survival; PFS = progression-free survival; rCBV = relative cerebral blood volume; ROC = receiver operating characteristic

Glioblastoma (GBM) is the most common primary malignant tumor in the adult brain. Although its prognosis remains poor (median survival, ~14.7 months), some patients with GBM show a distinct prognosis and response to chemoradiation. Previous studies have investigated the prognostic factors of GBM, including MR imaging and molecular biomarkers (O6-methylguanine-DNA methyltransferase [MGMT] promoter methylation). Previous studies have shown that high relative cerebral blood volume (rCBV; derived via dynamic susceptibility contrast-enhanced MR imaging) and high volume transfer constant (Ktrans; derived via dynamic contrast-enhanced [DCE] MR imaging) are associated with poor survival outcomes. However, rCBV and Ktrans measurements have reproducibility issues related to post-processing techniques, including normalization, model-based calculation, arterial input function, and software. In contrast, the initial area under the time-to–signal intensity curve (IAUC),
derived via DCE MR imaging, is a model-free parameter that does not require an arterial input function or a complicated model-based calculation and is highly reproducible.\textsuperscript{10,14} Because IAUC reflects both tumor perfusion and permeability,\textsuperscript{9,10,15} we hypothesized that IAUC might be useful for predicting the survival outcome in GBM. In this study, we used automatically calculated IAUC parameters to maximize reproducibility.

**MATERIALS AND METHODS**

Our institutional review board approved this retrospective study and waived the requirement for informed consent.

**Study Population**

We reviewed the records of 101 consecutive patients who underwent preoperative DCE MR imaging for newly diagnosed GBM from October 2010 to July 2014. The enrollment process is shown in On-line Fig 1. The inclusion criteria were pathologically confirmed GBM, no prior biopsy or treatment, and available MGMT promoter methylation status. The exclusion criteria were previous brain biopsy or treatments ($n = 6$), unavailable MGMT status ($n = 3$), and inadequate DCE image quality, leading to processing errors ($n = 4$). Thus, 88 patients were enrolled.

**Imaging**

A 3T MR imaging unit (Achieva; Philips Healthcare, Best, the Netherlands) and an 8-channel sensitivity encoding head coil were used for preoperative MR imaging. The protocol included T1-weighted turbo spin-echo images with inversion recovery (TR, 2000 ms; TE, 10 ms; TI, 1000 ms; FOV, 240 mm; section thickness, 5 mm; matrix, 256 × 256), T2-weighted turbo spin-echo (TR, 3000 ms; TE, 80 ms; FOV, 240 mm; section thickness, 5 mm; matrix, 256 × 256), and T2-weighted fluid-attenuated inversion recovery (TR, 10,000 ms; TE, 125 ms; TI, 2500 ms; FOV, 240 mm; section thickness, 5 mm; matrix, 256 × 256) images. 3D T1-weighted turbo field echo images (TR, 9.8 ms; TE, 4.6 ms; FOV, 240 mm; section thickness, 1 mm; matrix, 224 × 224) were acquired after the injection of gadolinium-based contrast (0.1 mL/kg of gadobutrol, Gadovist; Bayer Schering Pharma, Berlin, Germany). For DCE-MR imaging, 60 dynamic phases of DCE T1-weighted images were acquired with the following parameters: TR, 6.3 ms; TE, 3.1 ms; FOV, 240 mm; matrix, 192 × 192; section thickness, 3 mm; flip angle, 15°. After we acquired the fifth phase, gadolinium-based contrast (0.1 mL/kg of gadobutrol) was injected (rate, 3 mL/s). The total acquisition time for DCE-MR imaging was 6 minutes 18 seconds and the temporal resolution was approximately 6 seconds.

**Image Postprocessing and Analysis**

The ROIs were drawn by the consensus of 2 neuroradiologists (Y.S.C. and S.-K.L., with 4 and 20 years of experience in neuroradiology, respectively) who were blinded to the MGMT status and other clinical information, using a semiautomatic signal-intensity-threshold method to include the entire enhanced portion of the tumors on postcontrast T1-weighted images, while avoiding necrotic or cystic portions.

IAUC values were acquired by integrating the area under the time-to-signal-intensity curve, from 0 to 30 seconds (IAUC30) and 0 to 60 seconds (IAUC60) after contrast agent arrival on the basis of previous studies.\textsuperscript{10} IAUC values were automatically calculated without any user-dependent input, to maximize interrater reliability, through automatic determination of the time of contrast agent arrival for each pixel; the time of contrast agent arrival was defined as the initial time at which a signal increase of >5% of the maximum signal increase was observed during >2 serial phases. The combination of this automatic calculation of IAUC and semiautomatic ROI segmentation of enhancing tumor yields almost perfect reproducibility.\textsuperscript{16} The IAUC values were normalized to the maximum signal increase. A detailed description of the image processing is provided in the On-line Appendix. The mean values of IAUC30 (IAUC30\textsubscript{mean}), and IAUC60 (IAUC60\textsubscript{mean}) were extracted from the enhanced tumors. Images were analyzed with Medical Image Processing, Analysis, and Visualization (Version 7.0; National Institutes of Health, Bethesda, Maryland) and R software, Version 3.0.2 (R Foundation, http://www.r-project.org/).

**MGMT and Other Prognostic Parameters**

MGMT promoter methylation status was evaluated with a methylation-specific polymerase chain reaction, which was retrospectively recorded from pathologic reports. Because IAUC may be related to other genetic profiles, including isocitrate dehydrogenase 1 (IDH1) mutation status, epidermal growth factor receptor (EGFR) expression, and the presence of 1p/19q codeletion, the data regarding these profiles were also recorded as described in the On-line Appendix. The presence of oligodendroglial components was also recorded because the presence of these components is related to better prognosis.\textsuperscript{17} Furthermore, we recorded the following clinical factors: age, sex, Karnofsky performance status, extent of resection, and postoperative treatment, along with conventional MR imaging findings. The extent of resection was categorized as total, subtotal (<100% but ≥75% of gross tumor removal), or partial (<75% of gross tumor removal) resection/biopsy based on the intraoperative impressions of the surgeons (J.H.C., S.-G.K., and E.H.K., with 20, 16, and 10 years of experience in neurosurgery, respectively) and postoperative MR imaging. Postoperative treatment was categorized as concurrent chemoradiation with temozolomide, radiation therapy only, temozolomide-based chemotherapy only, and no postoperative treatment.

Conventional MR imaging findings included volume of enhanced tumor, degree of edema, presence of non-contrast-enhanced tumor (nCET), presence of deep white-matter inva­sion, and eloquent brain involvement, which are associated with prognosis.\textsuperscript{2,3} Conventional MR imaging findings were de-
Statistical Analysis
Overall survival (OS) duration was defined as the time from MR imaging–based diagnosis to death or the final follow-up in living cases. Progression-free survival (PFS) duration was defined as the time from MR imaging–based diagnosis to tumor progression, recurrence, death, or the final follow-up in cases in which no tumor progression or death was noted, according to the Response Assessment in Neuro-Oncology as described elsewhere. Univariate survival analyses for OS and PFS of the entire cohort were performed with Kaplan-Meier estimates and the log-rank test; the analyses included IAUC30_{mean}, IAUC60_{mean}, MGMT status, the presence of an oligodendroglial component, clinical factors (age, sex, Karnofsky performance status, extent of tumor resection, and postoperative treatment), and conventional MR imaging findings (enhanced tumor volume, edema, nCET, eloquent brain involvement, and deep white-matter invasion). Subgroup survival analyses of the methylated and unmethylated MGMT groups were performed with the IAUC parameters. If the variables were continuous or had >3 categories, they were dichotomized by the optimal cutoff values for OS and PFS that were calculated by using the “cutp” function of the “survMisc” R package, as defined by Contal and O’Quigley. This method determines the optimal cutoff value yielding the most significant prognostic differences between the 2 groups based on test scores from the Cox regression model.

The multivariate models were established with the significant factors of univariate analyses for OS and PFS in the unmethylated MGMT promoter group because the IAUC parameters were prognostic for the unmethylated MGMT group only, as shown later. In the multivariate models, all the variables that were continuous or had >3 categories were dichotomized, as in the univariate analyses. The multivariate models for OS and PFS were established from other prognostic factors only (model 0), and IAUC30_{mean} and IAUC60_{mean} were subsequently added as covariate parameters (IAUC30_{mean} in model 1 and IAUC60_{mean} in model 2). These multivariate models before (OS and PFS model 0) and after (OS and PFS models 1 and 2) adding IAUC parameters were explored and compared using 2 methods: 1) The diagnostic accuracies for 1.5-year OS and 1-year PFS prediction were assessed by receiver operative characteristic (ROC) analysis. Then, the reclassification improvement on adding the IAUC parameters was assessed using net reclassification improvement, as described by Pencina et al. The random survival forest method was used to determine the variable importance of the IAUC parameters in prognosis prediction. The forests consisted of 50,000 trees, having a minimum of 3 samples in a terminal node. The IAUC parameters and other prognostic factors were ranked by the relative importance scores, and the C-indices of the random survival forest models were calculated. Representative trees were constructed with the variables that showed the highest importance scores.

We compared the IAUC parameters according to MGMT status and the other genetic profiles using the Student t test, on the basis of the results of the Shapiro-Wilk test for normality. Statistical analyses were performed with Stata (Version 12.1; StataCorp, College Station, Texas) and R for Windows, Version 3.0.2. A P value <.05 was considered significant.

RESULTS
The clinical characteristics of the patients according to the MGMT status are summarized in On-line Table 1. None of the clinical characteristics, except for median OS and PFS, showed a significant difference between the methylated and unmethylated MGMT groups. The median OS and PFS were 442 days and 272 days, respectively, in the unmethylated MGMT group, whereas they were 790 days and 657 days, respectively, in the methylated MGMT group (log-rank test, P = .014 and .003 for OS and PFS, respectively). The other genetic characteristics of patients are summarized in On-line Table 2.

Univariate Survival Analyses
The results of the univariate analyses for OS and PFS in the entire cohort are summarized in On-line Table 3. The IAUC parameters were not significant predictors of OS and PFS in the entire cohort, though high IAUC parameters tended to be associated with poor OS (P = .090 for IAUC30_{mean}, P = .072 for IAUC60_{mean}). Unmethylated MGMT, 66 years of age or older, less than total resection, postoperative treatment other than concurrent chemoradiation with temozolomide, and enhanced tumor volume of ≥30.2 cm³ were associated with poor OS (P < .05 for all). Similarly, unmethylated MGMT, 70 years of age or older, less than total resection, postoperative treatment other than concurrent chemoradiation with temozolomide, and enhanced tumor volume of ≥30.2 cm³ were associated with poor PFS (P < .05 for all). Eloquent brain involvement was associated with poor OS (P = .033), but not with PFS. Sex, Karnofsky performance status, edema, nCET, and deep white-matter invasion were not associated with OS or PFS.

The results of the subgroup univariate analyses of the IAUC parameters in the methylated and unmethylated MGMT groups are summarized in Table 1 and Fig 1. High IAUC parameters were significant predictors of poor OS (P = .008 for IAUC30_{mean}, P = .015 for IAUC60_{mean}) and poor PFS (P = .010 for IAUC30_{mean}, P = .015 for IAUC60_{mean}) in the unmethylated MGMT group, but not in the methylated MGMT group. Therefore, we established multivariate models for the unmethylated MGMT group alone. Representative cases of GBMs with unmethylated MGMT promoter status are shown in Figs 2 and 3.
The results of the ROC analyses with the multivariate models for 1.5-year OS and 1-year PFS in the unmethylated MGMT group are summarized in Table 2. Among the 51 patients in the unmethylated MGMT group, 8 were excluded from the 1.5-year OS analysis and 1 was excluded from the 1-year PFS analysis because the follow-up duration was insufficient. Both OS and PFS model 0 included age, extent of resection, postoperative treatment, and enhanced tumor volume as covariates; eloquent brain involvement was included as a covariate in the OS model 0 only, according to the results of the univariate analyses. When the IAUC30mean or IAUC60mean were added in the OS model 0 and PFS model 0 as covariates, the diagnostic accuracies improved for 1.5-year OS prediction (OS models 1 and 2 area under the receiver operating characteristic curve [AUC], 0.86) and 1-year PFS prediction (PFS model 1 AUC, 0.76; PFS model 2 AUC, 0.74), compared with model 0 (OS model 0 AUC, 0.81; PFS model 0 AUC, 0.69). Reclassification improvement by adding IAUC parameters was statistically significant, as assessed by net reclassification improvement ($P < 0.005$ and $P < 0.018$ for OS models 1 and 2; $P < 0.020$ with PFS model 1), except in the case of PFS model 2 with IAUC60mean, which showed borderline significance ($P < 0.059$).

### ROC Analyses and Net Reclassification Improvement for Predicting 1.5-Year OS and 1-Year PFS

The results of the ROC analyses with the multivariate models for 1.5-year OS and 1-year PFS in the unmethylated MGMT group are summarized in Table 2. Among the 51 patients in the unmethylated MGMT group, 8 were excluded from the 1.5-year OS analysis and 1 was excluded from the 1-year PFS analysis because the follow-up duration was insufficient. Both OS and PFS model 0 included age, extent of resection, postoperative treatment, and enhanced tumor volume as covariates; eloquent brain involvement was included as a covariate in the OS model 0 only, according to the results of the univariate analyses. When the IAUC30mean or IAUC60mean were added in the OS model 0 and PFS model 0 as covariates, the diagnostic accuracies improved for 1.5-year OS prediction (OS models 1 and 2 area under the receiver operating characteristic curve [AUC], 0.86) and 1-year PFS prediction (PFS model 1 AUC, 0.76; PFS model 2 AUC, 0.74), compared with model 0 (OS model 0 AUC, 0.81; PFS model 0 AUC, 0.69). Reclassification improvement by adding IAUC parameters was statistically significant, as assessed by net reclassification improvement ($P < 0.005$ and $P < 0.018$ for OS models 1 and 2; $P < 0.020$ with PFS model 1), except in the case of PFS model 2 with IAUC60mean, which showed borderline significance ($P < 0.059$).

### Contribution of IAUC Parameters in the Multivariate Model of Random Survival Forest

Random survival forests for OS showed that IAUC parameters were the second most important variable, following the volume of...
enhancing tumor as the top variable and followed by postoperative treatment as the third most important variable. The variable importance of random survival forests is shown in On-line Figs 2 and 3. The relative importance was 41% for IAUC30mean in OS model 1 and 57% for IAUC60mean in OS model 2, assuming that the relative importance of volume of enhancing tumor was 100%.

Generation of a signal tree from this forest using the top 2 predictors (ie, volume of enhancing tumor and IAUC parameters) resulted in volume of enhancing tumor as the primary split and IAUC parameters as the secondary split for a group with a low volume of enhancing tumor (Fig 4). This splitting caused a significant difference in survival between the groups (P < .001 for trees with IAUC30mean and IAUC60mean, log-rank test), with a difference in mean survival of 354 days and 388 days between the high volume of enhancing tumors versus low volume of enhancing tumors with low IAUC30mean and IAUC60mean, respectively. The C-indices were higher after the IAUC parameters were added in OS models 1 and 2 (C-index = 0.67), compared with OS model 0 (C-index = 0.61).

Random survival forests for PFS showed that IAUC30mean in PFS model 1 was the most important variable, followed by extent of resection and volume of enhancing tumor as the second and third most important variables, respectively, whereas in PFS model 2, IAUC60mean was the third most important variable, following the extent of resection and volume of enhancing tumor as the first and second most important variables, respectively. The relative importance was 100% for IAUC30mean, and 94% for extent of resection in PFS model 1. Generation of a single tree with

Table 2: Reclassification improvement by adding IAUC parameters for predicting 1.5-year OS (n = 43) and 1-year PFS (n = 50) in the unmethylated MGMT group

| Survival Time/Models | AUC (95% CI) | NRI (95% CI) | P Value* |
|----------------------|-------------|-------------|----------|
| 1.5-year OS (n = 43) |             |             |          |
| Model 0 (ref)        | Age, extent of resection, postoperative treatment, volume of enhancing tumors, eloquent brain involvement | 0.81 (0.65–0.96) | – | – |
| Model 1              | IAUC30_mean | 0.86 (0.74–0.98) | 0.95 (0.61–1.29) | .005 |
| Model 2              | IAUC60_mean | 0.86 (0.72–0.99) | 0.81 (0.47–1.15) | .018 |
| 1-year PFS (n = 50)  |             |             |          |
| Model 0 (ref)        | Age, extent of resection, postoperative treatment, volume of enhancing tumors | 0.69 (0.54–0.84) | – | – |
| Model 1              | IAUC30_mean | 0.76 (0.62–0.90) | 0.69 (0.39–0.98) | .020 |
| Model 2              | IAUC60_mean | 0.74 (0.58–0.89) | 0.56 (0.26–0.85) | .059 |

Note: —NRI indicates net reclassification improvement; ref, reference.

* Calculated from category-less NRI.
However, the variability of rCBV ranges from the prognostic value of IAUC in GBM remains unreported that the reproducibility between the model-free parameters show high reproducibility and performance.

Hence, we hypothesized that IAUC would be useful for While 1 study similarly, IAUC values were paradoxically higher in the methylated MGMT group than in the unmethylated group.

The differences in the IAUC parameters according to the other genetic profiles are described in the On-line Appendix.

**DISCUSSION**

In this study, we found that higher IAUC values were predictors of poor OS and PFS in the unmethylated MGMT group and that IAUC parameters improved prognosis prediction of the unmethylated MGMT group, when added to other prognostic factors. Furthermore, IAUC values were paradoxically higher in the methylated MGMT group than in the unmethylated group.

rCBV and $K_{\text{trans}}$ values are indicators of tumor perfusion and permeability, respectively, and high values are predictors of worse prognosis. However, the variability of rCBV ranges from 24% to 67%, depending on the normalization method or software used for postprocessing. Similarly, $K_{\text{trans}}$ can vary according to the model selection, arterial input function, software, and prebolus T1 measurement. In contrast, IAUC is a model-free parameter that does not depend on model selection and arterial input function and is highly reproducible. Whether model-free parameters show high reproducibility and performance, compared with model-based parameters, was debated recently. While 1 study reported that mean $K_{\text{trans}}$ showed wider interobserver variability than model-free parameters, another study reported that the reproducibility between the model-free parameter and $K_{\text{trans}}$ did not differ significantly. In the latter study, the time of contrast agent arrival was manually defined to calculate the model-free parameters. In our study, the time of contrast agent arrival was automatically determined for each pixel; With this automatic determination of the time of contrast agent arrival, the IAUC has been reported to achieve almost perfect reproducibility.

Despite the limitation of IAUC regarding the lack of a clear biologic association, IAUC values are correlated with $K_{\text{trans}}$ and rCBV. Hence, we hypothesized that IAUC would be useful for prognosis prediction because it reflects both perfusion and permeability, which are complementarily associated with prognosis. Although previous studies have reported that model-free parameters may be useful for differentiation of radiation necrosis and tumor, the prognostic value of IAUC in GBM remains unclear. Moreover, to establish the practical role of an imaging biomarker, one must elucidate whether that imaging biomarker has prognostic value after other prognostic factors are considered in the usual clinical settings. We found that IAUC parameters have prognostic value even after clinical factors and conventional MR imaging findings are considered in the unmethylated MGMT group. Therefore, IAUC represents a useful imaging biomarker that has practicality and high reproducibility. Nevertheless, considering that IAUC parameters did not achieve the same improvement in PFS prediction as IAUC30$_{\text{mean}}$, the optimal time point of IAUC needs to be determined in future investigations.

The association between high IAUC values and poor prognosis in the unmethylated group is consistent with findings in previous studies, which have reported the association between high rCBV and $K_{\text{trans}}$ values and poor prognosis. However, in our study, the IAUC values were not prognostic when applied to the methylated MGMT group or the entire GBM cohort, and IAUC values were paradoxically higher in the methylated MGMT group than in the unmethylated group. Some studies have reported the correlation between MGMT status and rCBV and $K_{\text{trans}}$ values, with conflicting results; one study reported that rCBV values were
higher in GBM cases with unmethylated MGMT than in those with methylated MGMT. In contrast, another study reported that $K^{\text{trans}}$ values were higher in GBMs with methylated MGMT, a finding discordant with the previously reported association between high $K^{\text{trans}}$ values and poor prognosis, and the authors hypothesized that high permeability of the methylated MGMT group allowed better penetration of temozolomide, yielding better response and prognosis.

Although the underlying mechanism is unclear, our results may reflect IAUC values being more affected by permeability difference than perfusion difference between the methylated and unmethylated MGMT groups. Subsequently, a combination of these paradoxically high IAUC values of the methylated MGMT group and the effect of methylated MGMT as a good prognostic factor might have counterbalanced the association between high IAUC values and poor prognosis. This possibility is 1 explanation for our result that the IAUC values were not prognostic in the methylated MGMT group or the entire cohort. Similarly, a previous study reported that low $K^{\text{trans}}$ values were associated with poor prognosis in GBM, which is contradictory to the results of other studies on the association between high $K^{\text{trans}}$ values and poor prognosis. These conflicting results might be attributable to the complicated relationship among tumor permeability, MGMT status, and prognosis, as observed in our study. Future investigation with a larger cohort is necessary to analyze the relationship among tumor permeability, MGMT status, and prognosis, as observed in our study. Future investigation with a larger cohort is necessary to analyze the relationship among tumor permeability, MGMT status, and prognosis as observed in our study.

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CONCLUSIONS

Our study demonstrated that IAUC could be a useful prognostic imaging biomarker in patients with GBM with known MGMT status, improving the prognosis prediction of GBM with unmethylated MGMT promoter status. We propose that this complementary role of IAUC and MGMT status should be considered in future investigations.

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