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Reducing False-Positives in CT Perfusion Infarct Core Segmentation Using Contralateral Local Normalization

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ABSTRACT

BACKGROUND AND PURPOSE: The established global threshold of rCBF <30% for infarct core segmentation can lead to false-positives, as it does not account for the differences in blood flow between GM and WM and patient-individual factors, such as microangiopathy. To mitigate this problem, we suggest normalizing each voxel not only with a global reference value (ie, the median value of normally perfused tissue) but also with its local contralateral counterpart.

MATERIALS AND METHODS: We retrospectively enrolled 2830 CTP scans with suspected ischemic stroke, of which 335 showed obvious signs of microangiopathy. In addition to the conventional, global normalization, a local normalization was performed by dividing the rCBF maps with their mirrored and smoothed counterpart, which sets each voxel value in relation to the contralateral counterpart, intrinsically accounting for GM and WM differences and symmetric patient individual microangiopathy. Maps were visually assessed and core volumes were calculated for both methods.

RESULTS: Cases with obvious microangiopathy showed a strong reduction in false-positives by using local normalization (mean 14.7 mL versus mean 3.7 mL in cases with and without microangiopathy). On average, core volumes were slightly smaller, indicating an improved segmentation that was more robust against naturally low blood flow values in the deep WM.

CONCLUSIONS: The proposed method of local normalization can reduce overestimation of the infarct core, especially in the deep WM and in cases with obvious microangiopathy. False-positives in CTP infarct core segmentation might lead to less-than-optimal therapy decisions when not correctly interpreted. The proposed method might help mitigate this problem.

 $\label{eq:BBREVIATIONS: MA = microangiopathic alteration; NoMA = no microangiopathic alteration; rCBF = relative CBF; rCBV = relative CBV; T_{max} = maximum of the residue function$

n most centers, indication for therapy of acute stroke is based on multimodal CT imaging protocols that comprise noncontrast CT, CTA, and CTP imaging.¹ CTP is a powerful tool in stroke imaging with proved high accuracy for the detection of ischemic lesions and the delineation of irreversibly damaged tissue (the so-called infarct core) and the penumbra that represents tissue potentially salvageable if reperfused. CTP was successfully employed as a selection criterion for endovascular therapy in 2 important trials, showing benefit for patients with limited-size-

Indicates article with online supplemental data. http://dx.doi.org/10.3174/ajnr.A8111 predicted infarct core and substantial salvageable brain tissue.^{2,3} Therefore, current stroke guidelines require CTP in patients with extended and unclear time windows.⁴ Despite recent studies on thrombectomy in patients presenting with large ischemic core,^{5,6} estimation of the infarct core by CTP is of central importance for patient selection in the prolonged time window.^{2,4}

To approximate the infarct core and the penumbra, metrics based on the perfusion bolus dynamics are calculated and compared between ischemic and healthy tissue.⁷ In detail, the time to peak of the contrast agent bolus (T_{max} ; ie, the maximum of the residue function) is employed to assess a delay in perfusion. The T_{max} is given in absolute units of seconds, and the typically used threshold of $T_{max} > 6$ seconds represents a relevant delay.⁸ The CBF, on the other hand, cannot be easily quantified and is therefore typically normalized to a reference value. Typically, as reference, the median value of the hemisphere contralateral to the stroke side or the median CBF of the patient's normally perfused tissue identified as $T_{max} \leq 4$ seconds is used. The infarct core is obtained by thresholding this normalized relative map with rCBF <30%.^{1,9}

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This "global" normalization approach can easily be implemented and standardized but is potentially hampered by several factors: It does not account for the differences in perfusion dynamics in GM and WM¹⁰⁻¹² and patient-individual characteristics, such as microangiopathic WM alterations.¹³ Especially the latter constitutes a major challenge in the calculation of perfusion values as the WM gliotic tissue shows a delayed and reduced perfusion already without acute vessel occlusion present.^{11,12,14} Whereas in general, overestimation of ischemic core volume is uncommon,¹⁵ these and other limitations can lead to false-positives in CTP infarct core segmentation¹⁶⁻¹⁸ and thus to misleading therapy decisions when not correctly interpreted.¹⁹

To overcome this technical challenge and account for microangiopathy as an important co-pathology in the context of cerebrovascular disease, we propose a dual approach. In addition to normalizing CBF values in each voxel with a global reference value, we suggest an additional normalization with local regions from the contralateral counterpart when calculating relative perfusion metrics. This comprehensive method accounts for both global and local factors, enhancing the accuracy of the analysis.

MATERIALS AND METHODS

Participants

For this retrospective study, we included patients who received neuroimaging, including CTP due to a suspected stroke between 2014 and 2021. For data management and image processing, all images were exported and anonymized to a local instance of the imaging platform NORA (www.nora-imaging.org). We excluded all patients with nonischemic lesions, such as tumors or hemorrhages. Image data were individually assessed for quality and patients with substantial image artifacts (ie, due to movement) were excluded from further analysis. Image data and patient records were assessed for the presence of microangiopathic brain damage.²⁰ We identified patients with relevant WM microangiopathic alterations (MA, eg, Fazekas >1) and assigned them to the MA group, whereas patients without relevant WM lesions (Fazekas 0 and 1) were assigned to the NoMA group. To accomplish this, an experienced neuroradiologist evaluated noncontrast CT data in 5-mm axial reconstructions by using a soft kernel. Furthermore, we excluded cases with old and bilateral infarctions.

The study was approved by the Institutional Review Board (Ethics Committee–University of Freiburg; EK 20/1047) and carried out in accordance with the Declaration of Helsinki and its later amendments. Due to the retrospective nature of this study, the need for written informed consent was waived.

CT Imaging

CT scans were performed on a 128-detector row (Somatom Definition Flash; Siemens) or a 64-detector row multidetector CT scanner (Somatom Definition 64 AS; Siemens). CTP series were acquired in the axial scan mode with the following protocols for the Somatom Definition Flash: 80 kV, 180 mAs, collimation = 16×1.2 mm, no gantry tilt, tube rotation time = 0.3 seconds, toggle technique (5 blocks \times 0.3 = 1.5 seconds), z-coverage = 100 mm, section thickness = 5 mm, increment = 5 mm, 27 series every 1.5 seconds = 42 seconds scan time

(effective temporal resolution in the middle of the plane = 1.5 seconds, in the periphery = 2.6 seconds) after intravenous (16–18 G) injection of 40 mL of Imeron 400 (iopamidol; Bracco) + 30 mL of sodium chloride (NaCl) at a flow rate of 6 mL/seconds.

The parameters for Somatom Definition 64 AS were as follows: 80 kV, 180 mAs, collimation $=32 \times 1.2$ mm, no gantry tilt, tube rotation time = 0.3 seconds, toggle technique (5 blocks \times 0.3 seconds = 1.5 seconds), z-coverage = 90 mm, section thickness = 5 mm, increment = 3 mm, 30 series every 1.5 seconds =45 seconds scan time (effective temporal resolution in the middle of the plane = 1.5 seconds, in the periphery = 2.6 seconds) after intravenous (16–18 G) injection of 40 mL of Imeron 400 + 30 mL of NaCl at a flow rate of 6 mL/seconds.

Calculation of Perfusion Values

Perfusion scans were analyzed by using a research version of the software package VEOcore (VEObrain, www.veobrain.com). The software provides fully automated processing of perfusion scans, including motion correction, denoising, deconvolution by using Tikhonov regularization, and automated quality control. The output of the software is perfusion maps of relative CBF (rCBF), relative CBV (rCBV), and T_{max}, as well as segmentations of hypoperfusion, core and mismatch, and corresponding volumes in milliliters by using the commonly established thresholds of T_{max} > 6 seconds for delayed perfusion and rCBF <30% for the infarct core.^{2,21,22} Good agreement of VEOcore with other perfusion software has been demonstrated.^{23,24}

Our study employed 2 distinct normalization approaches for thresholding the rCBF maps. First, we adopted the conventional well-established global normalization method, wherein each voxel's value is normalized to the average value of the contralateral hemisphere.⁹ Mathematically, this normalization is achieved by dividing each voxel's value by a common factor (ie, the median value of normally perfused tissue defined as $T_{max} \leq 4$ seconds).²⁵ It is important to note that this process globally scales the rCBF map but retains the relative differences between voxels.

Additionally, we investigated the feasibility and potential of a more localized normalization approach (a schematic is provided in Fig 1). The concept behind this method is to normalize each voxel not to a global reference value but rather to its corresponding counterpart in the contralateral hemisphere. To achieve this, a transformation is required to map each voxel to its direct counterpart. This was obtained by nonrigid registration of the baseline image of the perfusion scan with a vertically mirrored copy of that image by using SPM12 (Wellcome Trust Center for Neuroimaging). When applying this transformation to the rCBF data array, a mirrored array can be obtained that directly relates each voxel to its contralateral counterpart. Normalization was then accomplished through voxelwise division of the original image (ie, the whole brain CBF map) by the mirrored one. However, several considerations were taken into account. Brain symmetry is rarely perfect, and registration inherently entails some degree of uncertainty. Moreover, division by noisy maps can introduce instabilities. Furthermore, very high flow values in the counterpart, particularly in large vessels, can result in very low values after division, potentially



FIG 1. Schematic of the workflow for global and local normalization of CT perfusion-derived CBF with subsequent thresholding to 30% to obtain a segmentation of the ischemic core. In the global approach, each voxel is normalized to the entire contralateral hemisphere, whereas in the local approach, a voxel is compared with a local region on the contralateral side. This leads to an equalization between GM and WM and thus a more distinct contrast in the perfusion maps.

Patient characteristics

					Mean Core Volume (mL)	
	n	Sex (Female)	Age (yr)	Stroke Evident	Global Normalization	Local Normalization
No microangiopathy	2830	49%	72.6 ± 15.2	1334 (47%)	41.0 ± 46.9	31.4 ± 41.1
Microangiopathy	335	59%	73.5 ± 17.1	175 (52%)	60.3 ± 49.0	33.6 ± 39.3

leading to false-positives. To address these issues, 2 precautionary measures were implemented before division: The mirrored image was clipped, meaning that any value exceeding 150% was capped at 150%, and the mirrored image then underwent smoothing with a Gaussian filter with a sigma of $12 \times 12 \times 4$ mm.

The local approach allows each voxel's value to be related to a local region of the contralateral hemisphere, inherently accommodating differences between GM and WM¹¹ and accounting for patient-specific microangiopathic changes, which typically manifest symmetrically.²⁶ With these locally normalized maps, a threshold of 30% can now be interpreted as 30% relative to a local contralateral region, rather than relative to an averaged reference value.

Finally, for both the globally and locally normalized rCBF maps, a threshold of relative rCBF <30% was applied. Subsequent morphologic image opening and closing operations were applied to remove speckle noise followed by the calculation of infarct volumes.²⁷ To avoid false-positives on the healthy hemisphere, segmentations were restricted to the affected hemisphere only, determined as the one with the greater median T_{max} .

Statistical Analyses

Statistical analyses were performed by using MATLAB (MathWorks). We generated correlation plots for both the NoMA and MA groups. The average changes in infarct volumes when comparing globally and locally normalized rCBF maps for both groups were assessed by relative means by using the linear

regression coefficient and by absolute means by using the mean differences and SDs between both approaches derived from Bland-Altman plots between both approaches.

Data Availability

Data and code are available from the authors upon reasonable request and approval by the local ethics committee.

RESULTS

Participants

We retrospectively enrolled 2830 CTP scans of patients with suspected ischemic stroke (52% women, mean age 73 \pm 15 years), of which 335 showed obvious signs of microangiopathy corresponding to the MA group. While age did not significantly differ between the MA and NoMA groups (P = .2), significantly more women had MA (P < .001). Further information is provided in the Table.

Perfusion Values before and after Local Normalization

Correlation and Bland-Altman plots of the calculated core volumes by using both approaches are presented in Figure 2. On average, all volumes were reduced upon applying the local normalization approach. These reductions were significantly higher in the MA group (mean difference -14.7 mL, SD 14.6 mL, regression slope 0.62) and less pronounced in the NoMA group (mean difference -3.7 mL, SD 6.7 mL, regression slope 0.82) (P < .001). Visual assessment of the MA group perfusion maps revealed in particular



FIG 2. Correlation plot (*A*) and Bland-Altman plot (*B*) comparing core volumes derived with global and local normalization for the NoMa and MA groups (*black* and *red*, respectively). *Dashed lines* indicate linear regression in the correlation plot, and 2 SDs in the Bland-Altman plot.

a considerable reduction of positively scored voxels in the deep WM. Figures 3 and 4 show exemplary comparisons for both NoMa and MA cases.

DISCUSSION

The proposed method of local normalization can significantly reduce the extent of false-positive voxels in infarct core segmentation, especially in the deep WM and in cases with obvious microangiopathy.

This is of considerable relevance for the application of CTP in clinical routine as perfusion imaging is an increasingly utilized component in the workup and management of patients with acute ischemic stroke, as the decision to use recanalizing therapy frequently relies on CTP results.⁴ False-positive voxels in perfusion imaging-derived calculations can thus errone-ously influence therapy decisions.

Our data suggest that especially microangiopathic changes in the deep WM constitute a relevant influence on the calculation of infarct core volumes. In the context of cerebrovascular disease, microangiopathic changes are a frequent co-pathology in patients with vascular disease,²⁸ so a correction of false-positive values caused by microangiopathy might be relevant in the presence of small vessel disease. This is corroborated by the fact that we noted a substantially higher reduction of core volumes in cases with microangiopathy versus without. Moreover, visual inspection of cases with small vessel disease revealed that the proposed correction method mainly removed false-positives in the deep WM, which is supported by the case in Figure 4.

However, in patients without obvious microangiopathy, the local normalization approach also resulted in a reduction of infarct volumes of \sim 3.7 mL, or 18% on average. Again, this was attributed to WM regions, where the blood flow is naturally lower compared with the GM.^{10,11,29-32} In contrast to the global normalization, the introduced local approach allowed for a more distinct calculation of the perfusion in the GM and WM, thus leading to fewer voxels identified as ischemic upon applying the threshold of rCBF <30%. Whether such voxels truly are false-positives, or whether the threshold of rCBF <30% needs to be adapted to better delineate the ischemic tissue and

avoid underestimation requires evaluation in future studies, including followup imaging.

Previously, the term "ghost infarct core" was introduced to characterize instances of false-positives in infarct core segmentation,¹⁶ and several underlying factors, such as poor collaterals, have been suggested.¹⁷⁻¹⁹ In our study, we have identified and analyzed an additional factor that warrants consideration.

Novel infarct segmentation techniques leveraging deep learning are rapidly advancing.³³⁻³⁷ These techniques not only perform with sufficient precision but also allow for a more multi-

parametric approach by simultaneously taking information from different raw data and derived features into account. Furthermore, some methods are able to account for contralateral asymmetries through innovative data augmentation methods, such as vertical image flipping.³³ Against this, we perceive our local normalization approach not as a competitor to deep learning but rather as a potentially valuable contribution to both conventional and deep learning–based postprocessing. It is important to recognize that the effectiveness of deep learning substantially hinges on preprocessing steps and the judicious selection of appropriate training data sets, and our work offers insights that can potentially aid in refining deep learning models in both aspects.

In principle, differential GM and WM perfusion or microangiopathic damage may not only impact rCBF, but also T_{max}^{13} . However, because T_{max} is directly quantified in units of seconds and thus does not require normalization, and GM versus WM differences are less pronounced in T_{max}^{11} we did not consider a local normalization for T_{max} in the context of this study.

For rCBV evaluation, on the other hand, a local contralateral normalization might also offer benefits regarding visual and quantitative scoring of collateralization, as illustrated in the Online Supplemental Data for exemplary cases. A more detailed analysis of rCBV was beyond the scope of the present study and might be part of future research.

A general limitation of our work is the fact that we do not have a consistent reference standard, for example, an MR imaging examination as close in time as possible or a follow-up MR imaging indicating the final infarct volume. Also, an underlying assumption of our approach is that microangiopathic changes are largely symmetrical, which may not be true in rare cases. Another limitation is associated with the "healthy" hemisphere. In cases involving old or bilateral infarctions, this might lead to biased results (Online Supplemental Data). This issue is also pertinent in the global approach, but more pronounced when using local normalizations. It is crucial in such instances to cross-reference with the globally normalized maps for a more comprehensive assessment. Moreover, the local normalization approach induced a rather unusual impression of the rCBF maps and might hamper the assessment of the "healthy" hemisphere as visible in the



FIG 3. Comparison of global (*left*) and local (*right*) normalization for CBF maps and threshold-based ischemic core segmentations in 3 cases without and with WM small vessel disease.

exemplary cases in which the image impression is more of a mirrored picture of the ischemic hemisphere and does not allow for an intrahemispheric assessment of rCBF values. Hence, the proposed approach of a local contralateral normalization must not replace the conventional rCBF calculation but rather supplement it in cases with suspected false-positives (eg, due to severe



FIG 4. Image data of an 83-year-old woman 5 hours after onset of a left MCA occlusion. CTP with global contralateral normalization (*A*) suggested an infarct core involving the centrum semiovale (*B*). In contrast, CTP with local contralateral normalization (*C*) did not display the deep WM as infarct core (*D*). The M1 segment was recanalized following one thrombectomy maneuver by using a 4 \times 40 mm Solitaire stent (*E*, *F*). Follow-up MR imaging shows distinct microangiopathy (*G*, FLAIR), but only scattered subacute centrum semiovale infarct (*H*, DWI).

microangiopathy). Additionally, our results should raise attention to the potential impact of microangiopathy on rCBF values.

CONCLUSIONS

A local normalization of CTP parameters in addition to the global normalization might reduce false-positives in CTP infarct core segmentation that could potentially lead to suboptimal therapy decisions.

Disclosure forms provided by the authors are available with the full text and PDF of this article at www.ajnr.org.

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