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Tumor cellularity beyond the visible in soft tissue sarcomas: Results of an ADC-based, single center, and preliminary radiomics study

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Purpose: Soft tissue sarcomas represent approximately 1% of all malignancies, and diagnostic radiology plays a significant role in the overall management of this rare group of tumors. Recently, quantitative imaging and, in particular, radiomics demonstrated to provide significant novel information, for instance, in terms of prognosis and grading. The aim of this study was to evaluate the prognostic role of radiomic variables extracted from apparent diffusion coefficient (ADC) maps collected at diagnosis in patients with soft tissue sarcomas in terms of overall survival and metastatic spread as well as to assess the relationship between radiomics and the tumor grade.

Methods: Patients with histologically proven soft tissue sarcomas treated in our tertiary center from 2016 to 2019 who underwent an Magnetic Resonance (MR) scan at diagnosis including diffusion-weighted imaging were included in this retrospective institution review board–approved study. Each primary lesion was segmented using the b50 images; the volumetric region of interest was then applied on the ADC map. A total of 33 radiomic features were extracted, and highly correlating features were selected by factor analysis. In the case of feature/s showing statistically significant results, the diagnostic accuracy was computed. The Spearman correlation coefficient was used to evaluate the relationship between the tumor grade and radiomic features selected by factor analysis. All analyses were performed applying p<0.05 as a significant level.

Results: A total of 36 patients matched the inclusion criteria (15 women; mean age 58.9 \pm 15 years old). The most frequent histotype was myxofibrosarcoma (16.6%), and most of the patients were affected by high-grade lesions (77.7%). Seven patients had pulmonary metastases, and, altogether, eight were deceased. Only the feature lmc1 turned out to be a predictor of metastatic spread (p=0.045 after Bonferroni correction) with 76.7% accuracy. The value -0.16 showed 73.3% sensitivity and 71.4% specificity, and patients with metastases showed lower values (mean lmc1 of metastatic patients -0.31). None of the examined variables was a predictor of the overall outcome (p>0.05, each). A moderate statistically significant correlation emerged only between lmc1 and the tumor grade (r=0.457, p=0.005).

Conclusions: In conclusion, the radiomic feature Imc1 acts as a predictor of metastatic spread in patients with soft tissue sarcomas and correlates with the tumor grade.

KEYWORDS

radiomics, soft tissue sarcoma, magnetic resonance, tumor grade, metastasis

Introduction

Soft tissue sarcomas are rare tumors representing approximately 1% of all malignancies, and diagnostic radiology plays a significant role not only at diagnosis and staging but also during follow-up (1-3). In particular, MR imaging, because of its intrinsic soft tissue contrast, is considered the main tool for investigating the primary site. Furthermore, this technique carries additional advantages due to the possibility to perform quantitative and functional analyses. In fact, the information about tumor perfusion, chemical composition, and cellularity can be easily assessed by dynamic contrast-enhanced (DCE) techniques, spectroscopy, and diffusion-weighted imaging (DWI), respectively (3-9). Nowadays, DWI is considered crucial in oncological imaging in general (10, 11). For soft tissue sarcomas, it has been demonstrated that it also contributes in distinguishing recurrences from postsurgical scars, improves specificity in defining tumor margin infiltration, and predicts the response to treatment (8-10). Despite these encouraging results, potential pitfalls and controversies must be addressed, especially for myxoid tumors due to their high mucin content. In fact, in this case, the distinction between benign and malignant lesions by DWI could be hampered (10).

The recent technical developments allowed radiologists to move further in their contribution to the overall diagnostic management of cancer patients as demonstrated by the increasing use of complex analyses such as histogram and radiomics. Indeed, these types of computations were demonstrated to be very useful for different types of tumors including soft tissue sarcomas, applying various sequences (12–15). For instance, Corino et al. already showed that the variables of first order (FOS) extracted from apparent diffusion coefficient (ADC) maps allow the distinction of the lesions of different grades. Similar conclusions were drawn by Xu and colleagues using T1w and T2w fat-sat imaging (15, 16). Using DCE-MRI at the baseline in patients with high-grade non-metastatic soft tissue sarcomas, Crombe et al. showed that radiomic variables have an important prognostic role (17). Lastly, Gao and colleagues did not predict the response to radiotherapy by DWI but better predicted the treatment effect score applying delta radiomics (18).

Despite this growing evidence, up to now, to the best of our knowledge, the prognostic value of radiomics regarding metastatic spread has not been fully investigated yet.

Thus, the aim of this study was to evaluate the prognostic role of radiomic variables extracted from ADC maps collected at diagnosis in patients with soft tissue sarcomas in terms of overall survival and metastatic spread as well as to assess the relationship between radiomics and the tumor grade.

Methods

Patients with histologically proven soft tissue sarcomas treated in our tertiary center from 2016 to 2019 who

underwent an MR scan at diagnosis by a 1.5 T scanner (Siemens Avanto 1.5T, Siemens Healthcare, Siemens, Erlangen, Germany) including an axial short tau inversion recovery–DWI sequence with 6 mm slice thickness and two b-values (i.e., b50 and b 800) were examined for this preliminary retrospective single-center institution review board–approved study. One radiologist with 12 years of experience in musculoskeletal imaging segmented each lesion along tumor margins using the b50 images because of the higher spatial resolution than the ADC map. The volumetric region of interest was then applied on the map (Figure 1). From each segmented volume, the 33 radiomic features of two classes were extracted: intensity-based features

(FOS) and texture features [gray-level co-occurrence matrix (GLCM as well as gray-level run length matrix (GLRLM)]. The segmentation and extraction of radiomics features have been performed by an open-source software (3D Slicer, www. slicer.org).

Statistical analysis

Descriptive statistics were applied for demographics, tumor histotype, site, grade, and metastatic spread. Factor analysis was applied to select highly correlating radiomic features. Then,



Representation of the segmentation process and data extraction from the primary lesion of patients with soft tissue sarcoma included in the study. In particular, it is hereby represented by a 76-year-old woman affected by a myxofibrosarcoma of the right lower leg. The axial b50 DWI images have been used for segmentation of the entire lesion. The volumetric segmentation has been then applied on the apparent diffusion coefficient (ADC) map and the extraction of 33 radiomic features performed. Factor analysis allowed the selection of five highly correlating features. stepwise regression analysis was used to evaluate if any of the selected radiomic variables had a predicting role on the overall outcome (dead/alive) and/or the metastatic spread. Moreover, the Bonferroni correction was applied to correct the statistical significance level for multiple tests. In the case of feature/s showing statistically significant results, the diagnostic accuracy was computed using ROC curves and the value/s with the highest Youden index were selected as a cut-off. The Spearman correlation coefficient was used to evaluate the relationship between the tumor grade and all radiomic features selected by factor analysis.

To evaluate the robustness of the proposed method, all segmentations and data extraction were repeated by a second reader with 4 years of experience in oncological imaging and the intraclass correlation coefficient (ICC) of the variables highly correlating at factor analysis were computed. ICC values >.750 were considered excellent (19).

All statistical analyses were performed with SPSS (IBM SPSS Statistics version 27, IBM Armonk, NY, USA), applying p<0.05 as a significant level.

Results

From an overall amount of 80 cases treated in our center, 36 patients matched the inclusion criteria (15 women; mean age 58.9 ± 15 years old) and were examined. The characteristics of the examined population are summarized in Table 1. The most frequent histotype was myxofibrosarcoma (six patients, 16.6%), and most of the patients were affected by high-grade lesions (i.e., 28 had grade III lesions, 77.7%). Seven patients had pulmonary metastases, and, altogether, eight were deceased. On average, the survival was of 56.9 ± 22 months. Most of the lesions affected the lower limbs (29, 80.5%).

Factor analysis allowed the extraction of five highly correlating variables: three of first order (kurtosis, skewness, and uniformity) and two of second order (informational measure of correlation (Imc1) and high gray-level run emphasis).

Only the feature Imc1 turned out to be a predictor of metastatic spread (p=0.045 after Bonferroni correction) with 76.7% accuracy (Figure 2). The value -0.16 showed 73.3% sensitivity and 71.4% specificity, and patients with metastases showed lower values (mean Imc1 of metastatic patients -0.31). None of the examined variables was a predictor of the overall outcome (p>0.05 each).

A moderate statistically significant correlation emerged only between Imc1 and the tumor grade (r=0.457, p=0.005) (Figure 3).

The proposed method of the segmentation and extraction of radiomic features showed high reproducibility: kurtosis ICC = .870 [95% CI, .748 - .933], skewness ICC = .765 [95% CI, .748 - .933]

TABLE 1 Characteristics of the examined population.

Gender(female/male)	14/22
Age(years)	58.9 ± 15 (mean \pm SD)(range 18–82)
Histotype	6 myxofibrosarcoma
	6 undifferentiated pleomorphic sarcoma
	5 leiomyosarcoma
	4 myxoid liposarcoma
	3 aggressive fibromatosis
	2 synovial sarcoma
	2 solitary fibrous tumor
	2 liposarcoma
	2 round cell liposarcoma
	1 angiosarcoma
	1 hemangioendothelioma
	1 malignant peripheral nerve sheath tumor
	1 spindle cell sarcoma
Site	29 lower limbs
	3 upper limbs
	2 pelvis
	2 thorax
	1 neck
Grade	3 grade I
	5 grade II
	28 grade III
Metastatic(yes/no)	7/29
Deceased(yes/no)	8/28
Survival in months(median)	55

CI,.544 -.879], uniformity ICC = .814 [95% CI,.640 -.904], Imc1 ICC = .786 [95%CI,.584-.890], and high gray-level run emphasis ICC=.842 [95% CI,.692 -.918].

Discussion

To the best of our knowledge, this is one of the largest studies investigating the role of radiomic features extracted from the ADC maps of patients with soft tissue sarcomas at diagnosis demonstrating that the Imc1 is a predictor of metastatic spread and correlates with the tumor grade. Similar results were obtained using T1w and T2w fat-sat sequences by other groups. For instance, Tian and colleagues, applying a machine learning model, and Vallieres et al., associating the abovementioned MR sequences with positron emission tomography (PET) information, obtained an early detection of pulmonary metastases (13, 20). This last evidence, together with other previous promising results provided by PET/MR-based histogram analyses and texture features, suggests that the role of hybrid imaging should be further assessed, especially about



the possibility to simultaneously collect data regarding metabolic activity and functional information (12, 20).

Regarding, in particular, the Imc1 feature, it quantifies the complexity of the texture and its importance has been previously demonstrated in several studies on different types of cancer (21-24). For instance, our results are in line with those of Liao and colleagues who identified more negative values in patients with brain metastasis due to non-small lung cancer with poor local tumor control, thus suggesting that tumors with high intralesion heterogeneity might be associated with a worse clinical course, in our study represented by the occurrence of metastasis. Moreover, together with other features, it was part of a "radiomic signature" indicating patients with ground glass nodules at risk for invasive adenocarcinoma (25). The feature Imc1 may even be considered a novel biomarker of metastatic spread in patients with soft tissue sarcomas considering that it showed an excellent repeatability in our dataset and strong robustness in a previously published computed tomography-based computational model (26). Nevertheless, it should be considered that further studies on a larger population are needed to fully assess the hereby presented evidence; since, for instance, Corino et al., in a group of 19 patients with soft tissue sarcomas, showed that only the features of the first order were the best classifier of the tumor grade (15).

For the overall role of radiomics for this heterogeneous group of tumors, as previously underlined, different MR

sequences have been used with very promising results (17, 18) and probably not only more comparisons among sequences are necessary (27), but it should also be assessed if a model combining information deriving from multiple sequences could provide additional and more robust results.

In contrast to part of the literature, in our study, none of the selected radiomic features turned out to be a predictor of the overall outcome. This discrepancy might be due to the heterogeneity of our sample in terms of the histotype and grade. In fact, radiomics acted as prognostic factor in selected groups like patients with myxoid and liposarcomas only and patients with high-grade lesions (28–30). Moreover, in the above-mentioned studies, other types of sequences (i.e., T1w, T2 fat-sat, or DCE-MRI) than DWI have also been used. We strongly encourage further investigation in this direction to also gain new knowledge about the potential prognostic role of this sequence in the specific subgroups of patients.

This study is affected by several limits. First, as noted above, the sample size and the large variety of histotypes did not allow to investigate the predictive role of this technique on the type of tumor and the relationship between different histotypes and radiomic features. Moreover, the small population also hampered the subdivision of our sample in test and validation sets. Certainly, the rarity of the disease contributed to these flaws and multicenter studies are necessary to fill these voids and provide further insights into the role of radiomics for these tumors.



Then, the lower spatial resolution of DWI in contrast to conventional imaging should be considered. Nevertheless, as previously mentioned, analyses showed a high repeatability and several *ex vivo* and *in vivo* studies demonstrated a high reliability of radiomics using this type of sequence (31, 32).

In conclusion, the radiomic feature Imc1 acts as a predictor of metastatic spread in patients with soft tissue sarcomas and correlates with the tumor grade. Further studies on a larger sample and including delta-radiomics analyses at follow-up are expected to provide new insights on the potential impact of this evidence on the therapeutic and overall management of this rare group of tumors.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

Ethics statement

This study was reviewed and approved by the Ethics Committee of the Veneto Institute of Oncology (Approval No. 16336/21 CESC-IOV) on 1 September 2021.

Author contributions

All authors significantly contributed to this study and approved the final version of the manuscript. GC: study design, data analysis, and manuscript drafting. FG: data collection and review and editing of the manuscript. DP: data collection and review and editing of the manuscript. MS: data interpretation and review and editing of the manuscript. BA: assisted in data collection and review and editing of the manuscript. RM: study design and review and editing of the manuscript. SR: review and editing of the manuscript. SR: review and editing of the manuscript and study coordination. All authors contributed to the article and approved the submitted version.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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