

# Three-Dimensional CT Texture Analysis to Differentiate Colorectal Signet-Ring Cell Carcinoma and Adenocarcinoma

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Yali Yue \*  
Feixiang Hu\*  
Tingdan Hu  
Yiqun Sun  
Tong Tong  
Yajia Gu

Department of Radiology, Fudan  
University Shanghai Cancer Center,  
Shanghai 200032, People's Republic of  
China

\*These authors contributed equally to  
this work

**Purpose:** The objective of this research was to validate the diagnostic value of three-dimensional texture parameters and clinical characteristics in the differentiation of colorectal signet-ring cell carcinoma (SRCC) and adenocarcinoma (AC).

**Methods:** We retrospectively analyzed data from 102 patients with SRCC or AC confirmed by pathology, including 51 SRCC (from January 2015 to July 2019) and 51 AC patients (from January 2019 to July 2019). CT findings and clinical data, including age, gender, clinical symptoms, serological biomarkers, tumor size, and tumor location, were compared between SRCC and AC. CT texture features were quantified on portal phase images using three-dimensional analysis. A list of texture parameters was generated with MaZda software for the classification of tumors. The texture features, clinical data and CT findings were statistically analyzed for the discrimination ability of SRCC and AC, and the potential predictive parameters that may be used to differentiate the two groups were subsequently tested using the least absolute shrinkage and selection operator (LASSO) and logistic regression analyses. The receiver operating characteristic curve (ROC) provided a range of values for establishing the cutoff value, as well as the sensitivity and specificity of prediction for each significant variable.

**Results:** SRCC occurred more often in men than AC did (80.39% vs 49.02%,  $P < 0.01$ ). The patients were younger in the SRCC group than in the AC group, without a statistically significant difference (55.84 vs 59.20 years,  $P = 0.216$ ). There were no significant differences in the clinical symptoms, tumor size, or tumor location between the two groups ( $P=0.505$ ,  $P=0.19$ ,  $P=0.843$ , respectively). The elevation of serological biomarker CA724 was more common in SRCC than in AC ( $P < 0.001$ ). Perc.01%3D, Perc.10%3D and s(1,0,0) SumAverg were lower in the SRCC group than in the AC group during the portal phase, with the areas under curve (AUCs) of 0.892–0.929, sensitivity of 76.5–84.3% and specificity of 88.2–96.1%. In the differentiation between SRCC and AC, the 1-NN minimal classification error (MCR) was 29.41%.

**Conclusion:** Three-dimensional texture parameters, including Perc.01%3D, Perc.10%3D and s(1,0,0) SumAverg, exhibited a favorable discriminatory ability to distinguish SRCC from AC.

**Keywords:** computed tomography, colorectal signet-ring cell carcinoma, adenocarcinoma, texture analysis, three-dimensional

Correspondence: Yajia Gu; Tong Tong  
Department of Radiology, Fudan  
University Shanghai Cancer Center,  
No. 270 Dongan Road, Shanghai  
200032, People's Republic of China  
Tel +8618121299466;  
+8617349786369  
Fax +862164174774  
Email cjr.guyajia@vip.163.com;  
t983352@126.com

## Introduction

Colorectal cancer (CRC) is a major public health issue with a high prevalence of cancer-related morbidity and mortality, being the third most common malignancy worldwide.<sup>1</sup> Most colorectal cancers represent conventional adenocarcinomas

(ACs). Relatively rare pathological subtypes contain signet-ring cell carcinoma (SRCC) and mucinous adenocarcinoma (MAC). SRCC is a rare subtype of CRC, observed in approximately 1% of patients with colorectal cancer.<sup>2</sup> SRCC presents with greater than 50% of tumor cells showing abundant intracellular mucin, resulting in the eccentric displacement of the nuclei.<sup>3</sup> SRCC of the colorectum was reported for the first time in 1951.<sup>4</sup> Detecting the expression of various histological subtypes has important significance in accurate diagnosis, the judgment of biological behavior, and prognosis prediction for patients with colorectal cancer. Detection of histological subtypes may be a promising approach to guide clinical management.

Compared to other histological subtypes, SRCC has been linked to a higher risk of death and poorer prognosis. SRCC tends to present with more advanced stages mainly due to its late development of clinical symptoms.<sup>5</sup> Dismal prognosis after potentially curative surgery for colorectal SRCC remains a significant clinical problem compared with the prognosis for AC. When SRCC is clearly diagnosed, wider surgical resection and systemic preoperative radiotherapeutic treatment should be performed.<sup>6</sup> Therefore, correctly identifying SRCC and AC is important for clinical decision-making.<sup>7</sup>

Texture analysis, a potential noninvasive imaging biomarker, is a useful technique to assess the heterogeneity in routinely acquired images. To date, texture analysis has been widely applied in many kinds of tumors, including colorectal cancer, lung cancer, hepatic metastasis of colon cancer, esophageal cancer, etc., as a promising imaging biomarker of tumor treatment response and disease prognosis.<sup>8–12</sup> According to recently published reports, texture analysis of computed tomography (CT) was helpful for discriminating primary colon cancer from hepatic metastases and assessing survivability and chemotherapy response in patients with primary CRC.<sup>8,13</sup>

Texture analysis refers to the quantitative technique used to evaluate the characterizations of an ROI in a group of images. The proposed method is characterized by analyzing the distribution and relationship of pixel or voxel-gray levels, which may provide a measure of tumor heterogeneity.<sup>14,15</sup>

Thus far, no report on the differences in texture features between SRCC and AC has been published. Because of the poor tolerance of tumor patients, texture analysis is superior to invasive examinations such as colonoscopy and biopsy. Therefore, we aim to investigate the performance

and diagnostic value of three-dimensional texture analysis in differentiating SRCC from AC on CT images.

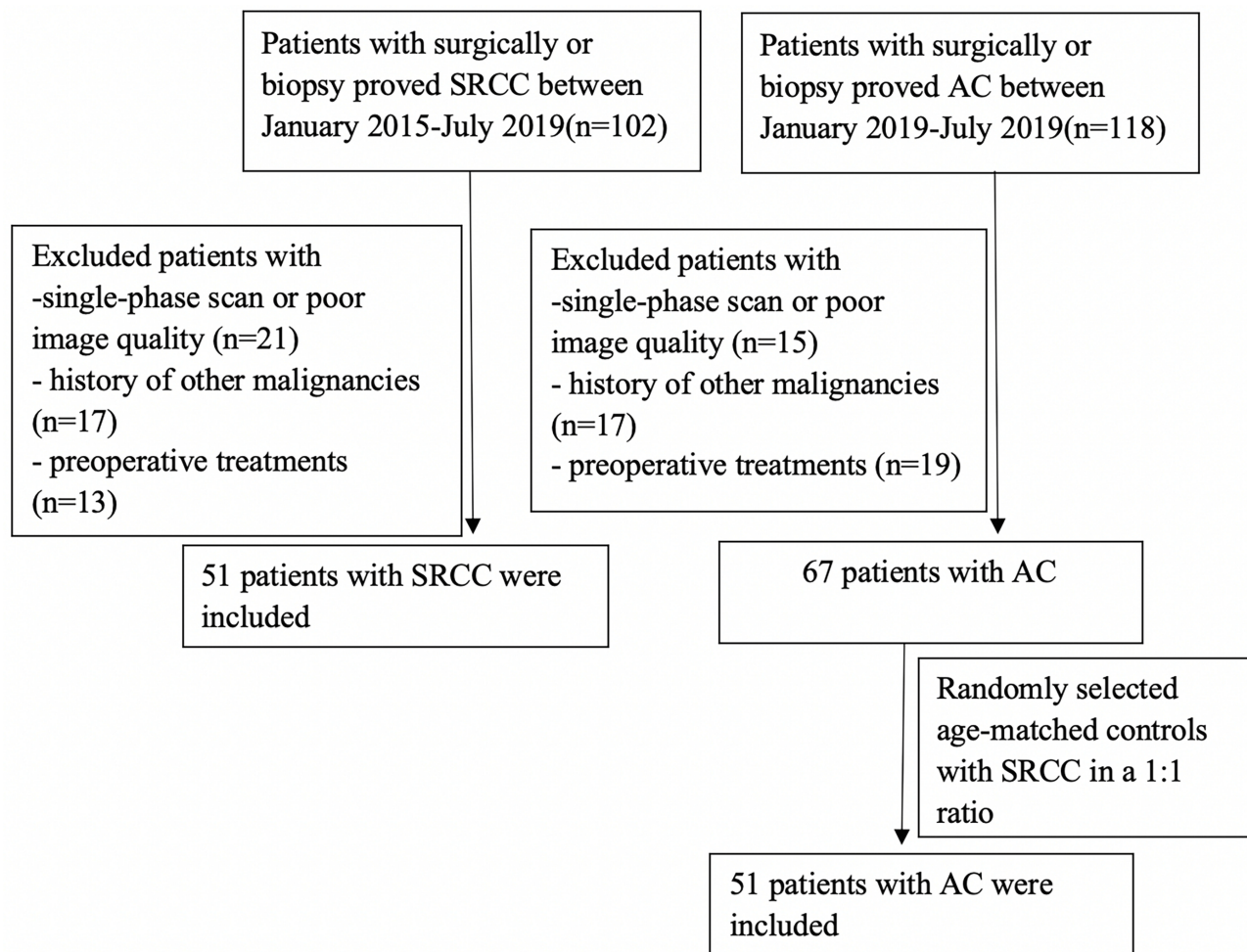
## Materials and Methods

### Patient Population

We identified 102 patients with pathologically diagnosed SRCCs between January 2015 and July 2019. 21 patients were excluded due to single-phase examination or poor image quality, 17 patients were excluded because of the history of other malignancies and 13 patients with preoperative treatment were excluded. Between January 2019 and July 2019, 118 patients, who were pathologically diagnosed with ACs enrolled in this study. 17 patients with a history of other malignancies were excluded. 21 patients were excluded due to a single-phase examination or poor image quality and 19 patients were excluded due to preoperative treatment. ACs were randomly selected age-matched controls with SRCC in a 1:1 ratio. Finally, 51 patients with AC and 51 patients with SRCC were included in our study. The inclusion criteria were as follows: 1) all diagnosis cases were verified by surgical resection or biopsy as AC or SRCC; 2) clinical features available; and 3) comprehensive CT data and medical history information. The following are the exclusion criteria: 1) single-phase scan or image quality does not meet the diagnostic requirements; 2) history of other malignancies within 5 years; 3) preoperative chemotherapy, radiation therapy or other local treatments. Finally, 102 patients (51 with AC and 51 with SRCC) were included in our study (Figure 1). This study was approved by the medical ethics committee of Fudan university cancer hospital. Written informed consents were obtained from patients before enrolling in the study. We also have complied with the World Medical Association Declaration of Helsinki regarding the ethical conduct of research involving human subjects.

### CT Examination

Patients underwent a multidetector CT examination (Somatom Definition, Siemens, Erlangen, Germany) in a head-first supine position following a standard operating procedure. The patients fasted for 8–12 hrs prior to CT scanning. All patients received plain, arterial, and portal venous phase CT scanning. Standard acquisition settings were as follows: automatic tube current modulation 160–251 mAs; tube voltage 120 kVp; slice thickness was set to 5.0 mm; helical pitch 1; and the reconstruction slice thickness was set to 1.25 mm for diagnostic reading. Iodohydrin (300 gI/mL) was administered at a dose of 1.5 mL/kg during contrast-enhanced CT scan.



**Figure 1** Flow diagram of patient inclusion.

**Abbreviations:** SRCC, signet-ring cell carcinoma; AC, adenocarcinoma.

A bolus-triggered technique was used to determine the scanning time, which was 30 s after the abdominal aorta reached a 100 HU triggering threshold, to perform portal venous phase axial images.

## CT Image Analysis

All images were independently reviewed by two radiologists (with 3 and 13 years of clinical experience in gastrointestinal CT, respectively) who specialized in gastrointestinal imaging. Readers were blinded to the pathological findings of each case. The imaging parameters of tumor location and size were reviewed for each patient by independent radiologists on the portal venous phase images. On the multiplanar reconstruction, the size was calculated by selecting the long diameter.

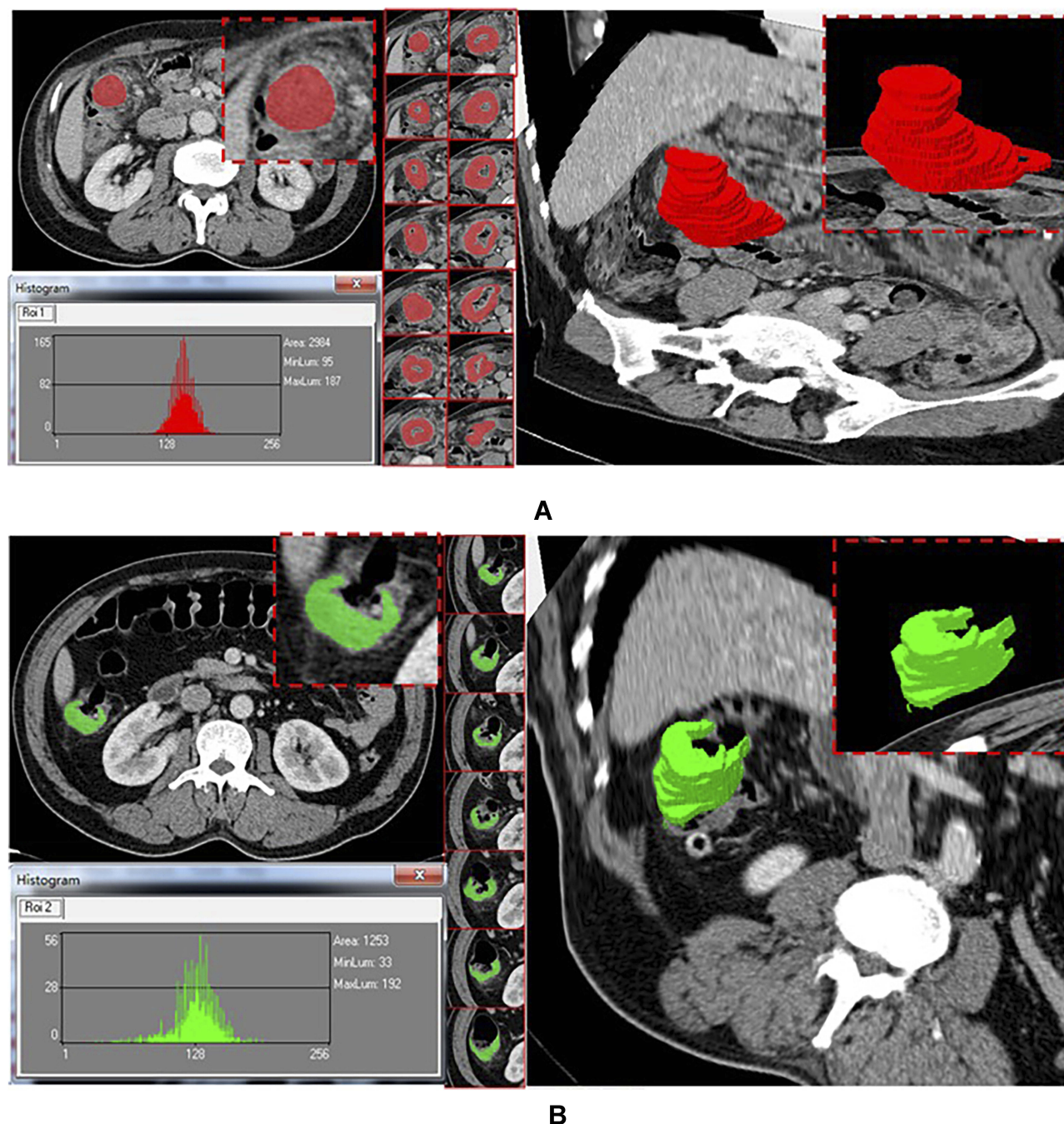
## CT Texture Analysis

MaZda software (Version 4.6, Instytut Elektroniki, available at <http://www.eletel.p.lodz.pl/mazda/>) was used to

delineate morphological tumor volumes in 3D on images. Images were stored in the Picture Archiving and Communication Systems (PACS) and imported into the software for analysis. CT images of the portal venous phase of all patients were ROI selected by consensus of the two attending radiologists. ROIs were manually drawn along the outline of the lesion on each consecutive slice to contain the entire tumor volume, avoiding air, peripheral vessels, and necrotic tissue and other irrelevant components during the delineation process (Figure 2). After the tumor segmentation was completed, the VOI (volume of interest) images were imported into MaZda software for feature extraction and analysis.

The following aspects were the main ways to extract texture parameters: histogram (information concerning the pixel intensities rather than the spatial relationship between gray values); gradient (information regarding the intensity distribution of an image); run-length matrix (adjacent or





**Figure 2** Texture analysis software program. **(A)** 67-year-old man, AC; **(B)** 59-year-old man, SRCC.

consecutive pixels/voxels runs with the specified gray-level values in a given direction); the cooccurrence matrix (information concerning changes in the gray-level value distribution of pixel pairs with increasing distance); and autoregressive model (texture description based on statistical correlations among multiple pixels). MaZda used the following criteria: the Fisher coefficient [Fisher], combination of classification error and correlation coefficient

[probability of classification error (POE) + average correlation coefficients (ACC)], mutual information and a selection of optimal feature subsets with minimal classification error (MCR) of the 1-nearest neighbor (1-NN) classifier. The first 10 texture features were chosen to constitute a feature subcategory for the subsequent discriminant in each discriminant group. Next, the linear discriminant analysis was used to identify the optimal discriminant

features for classification. Based on the feature subcategory extracted formerly, the 1-nearest neighbor (1-NN) classifier analysis was used to calculate minimal classification error (MCR).  $MCR (\%) = \text{total number of misdiagnosed cases} / \text{total number of cases in the differential group} \times 100\%$ . According to MCR, the discrimination ability was divided into five grades:  $MCR < 10\%$  is excellent,  $10\% < MCR < 20\%$  is good,  $20\% < MCR < 30\%$  is medium,  $30\% < MCR < 40\%$  is general, and  $MCR > 40\%$  is poor.

## Statistical Analysis

Mazda software can only filter the first 10 texture parameters, while the minimum absolute contraction selection operator (LASSO) algorithm can filter the most distinguishable predictive radiomics features. Single-factor logistic regression analysis was used to test the texture parameters and clinical characteristics. The potential predictive parameters to differentiate SRCC from AC were subsequently tested using multiple-factor logistic regression analysis. The diagnostic efficacy to differentiate SRCC from AC was assessed by ROC curve analysis. The data were analyzed using SPSS software (version 19, SPSS Inc., Chicago, IL, USA). Use R software (version 3.3.3; <http://www.Rproject.org>) for statistical analysis.  $P < 0.05$  was considered to indicate a statistically significant difference for the two groups.

## Results

The patient characteristics are listed in Table 1. The patients in the SRCC group were more likely to be men than those in the AC group (80.39% vs 49.02%,  $P < 0.05$ ). The patients in the group with SRCC were younger than those with AC were, without a statistically significant difference (55.84 vs 59.20 years,  $P = 0.216$ ). The clinical symptoms, tumor size and tumor location between the two groups showed no significant difference ( $P = 0.505$ ,  $P = 0.19$ ,  $P = 0.843$ , respectively). The elevation of serological biomarker CA724 was more common in SRCC than in AC ( $P < 0.001$ ). No differences were observed in elevated serological biomarkers, including CEA, CA199, CA125, CA50 and CA242, between the SRCC and AC groups ( $P > 0.05$ ).

The LASSO process is shown in Figure 3. Perc.01%3D, Perc.10%3D and s(1,0,0) SumAverg were chosen by Lasso. The chosen parameters were predominantly extracted from the histogram, gradient and cooccurrence matrix. Table 2 summarizes the area under the curve (AUC), specificity and sensitivity, and 95% confidence intervals were determined for clinical and texture features, which could be used to distinguish SRCC from AC. CT texture analysis has good

**Table 1** Clinical Features and CT Texture of SRCC and AC

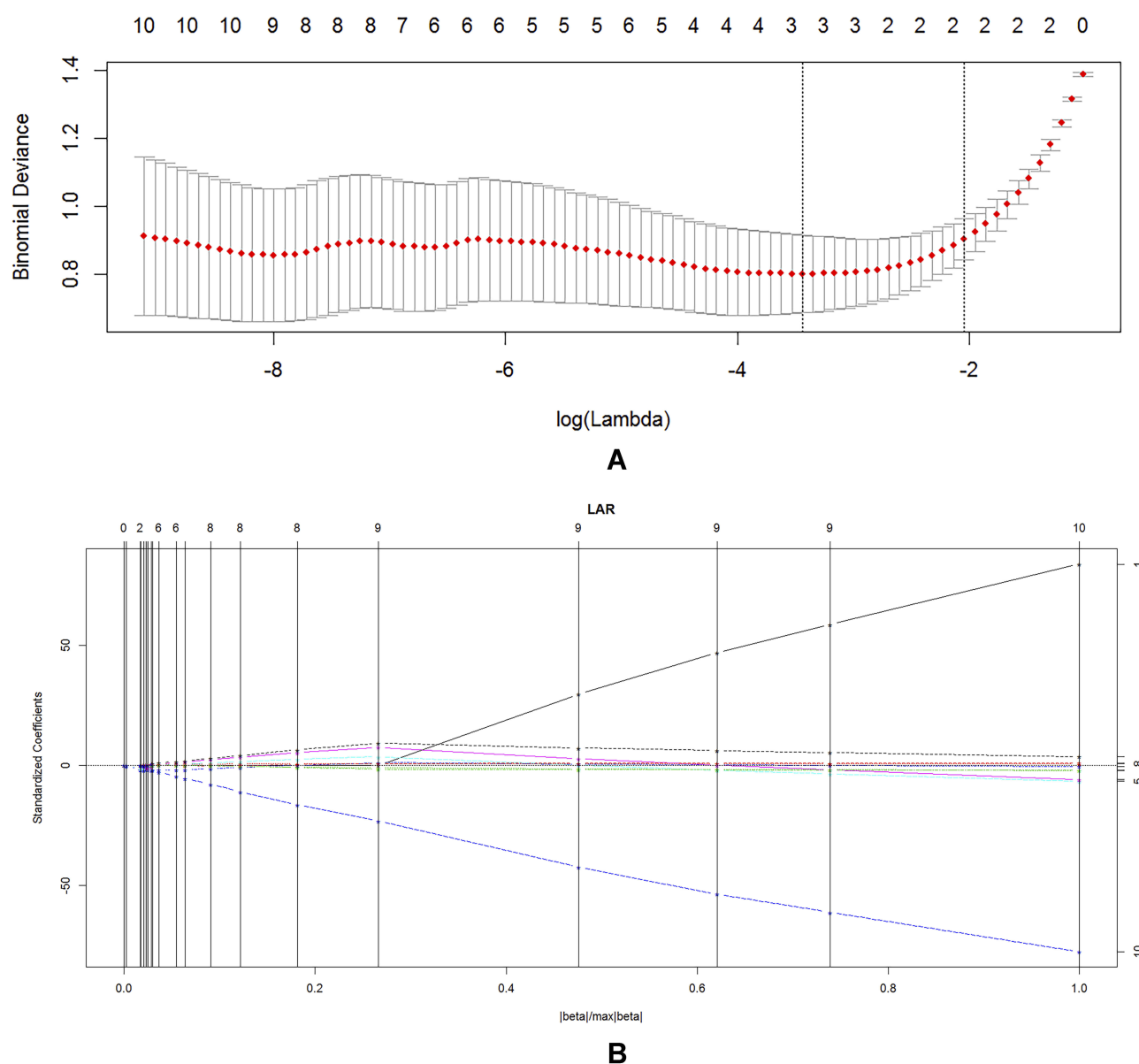
Characteristics	AC (n=51)	SRCC (n=51)	P Values
Age (years)	59.20±12.10	55.84±14.96	0.216
Gender			0.001
Male	25(49.02%)	41(80.39%)	
Female	26(50.98%)	10(19.61%)	
Size (mm)	48.81±19.31	42.49±12.64	0.19
Location			0.843
Left hemicolon	24(47.06%)	25(49.02%)	
Right hemicolon	27(52.94%)	26(50.98%)	
Clinical symptoms			0.505
Weight loss	6(11.76%)	4(7.84%)	
Asymptomatic	45(88.24%)	47(92.16%)	
CEA increase	22	23	0.842
CA199	12	13	0.818
CA125	2	6	0.269
CA724	4	29	<0.001
CA50	7	5	0.539
CA242	9	10	0.799

sensitivity and specificity in distinguishing SRCC from AC, which ranged from 76.5% to 84.3% and from 88.2% to 96.1%, respectively. The ROC analysis results are presented in Figure 4. The AUC ranged from 0.892 to 0.929. In the differentiation between SRCC and AC, the 1-NN minimal classification error (MCR) was 29.41% (Figure 5).

## Discussion

SRCC is a rare subtype of CRC, and patients with SRCC tend to present with a poorer prognosis than those with AC do.<sup>16</sup> Different histological subtypes of colorectal cancer have different biological behaviors, and thus the treatment response is not the same.<sup>17,18</sup> If the pathological subtypes can be predicted before treatment, it is expected to treat colorectal cancer with different biological behaviors by selecting a reasonable individualized treatment strategy.

CT texture analysis has been widely applied as a promising imaging biomarker in colorectal cancer.<sup>19</sup> Early studies have reported the application value of CT texture analysis in forecasting pathological characteristics, treatment response and patient prognosis.<sup>8,20,21</sup> However, no studies have been investigated into the texture analysis of SRCC alone or the texture comparison between SRCC and AC. In this study, we found that CT



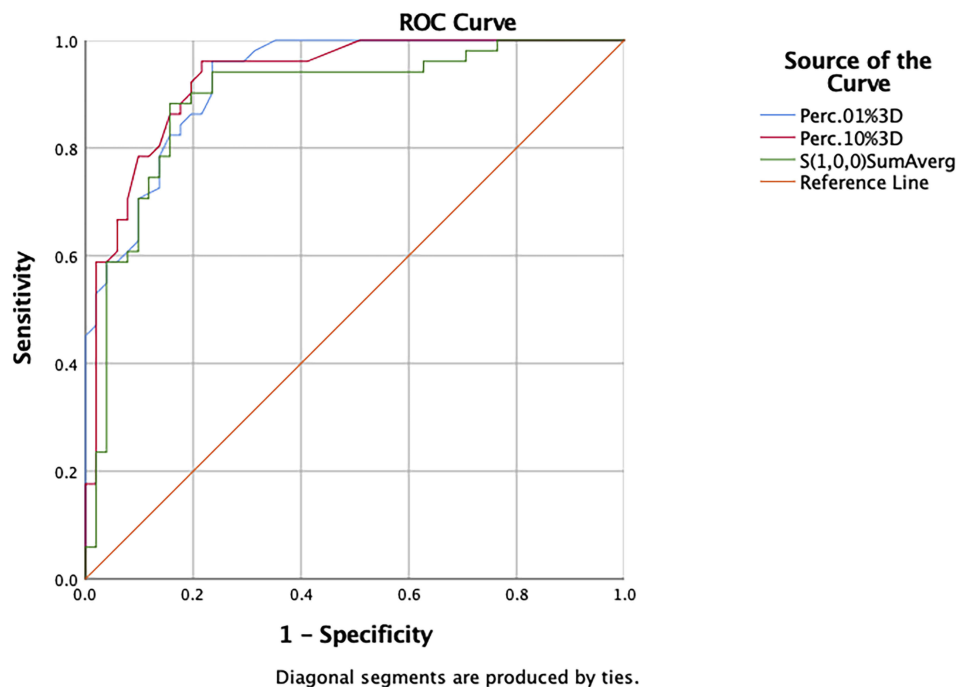
**Figure 3** Radiomics feature selection using the LASSO regression. LASSO, least absolute shrinkage and selection operator. **(A)** Tuning parameter ( $\lambda$ ) selection in the LASSO logistic model. The binomial deviance curve was generated vs  $\log(\lambda)$ . The minimum criteria for tenfold cross-validation were applied to  $\lambda$  selection. The optimal values of the LASSO tuning parameter ( $\lambda$ ) are indicated by the dotted vertical lines. **(B)** The vertical line corresponds to the number of iterations in lasso, and the independent variable nonzero coefficients are selected.

texture analysis obtained better performance in terms of differential diagnosis. Although CT texture analysis cannot replace pathological assessment, confirming tumor

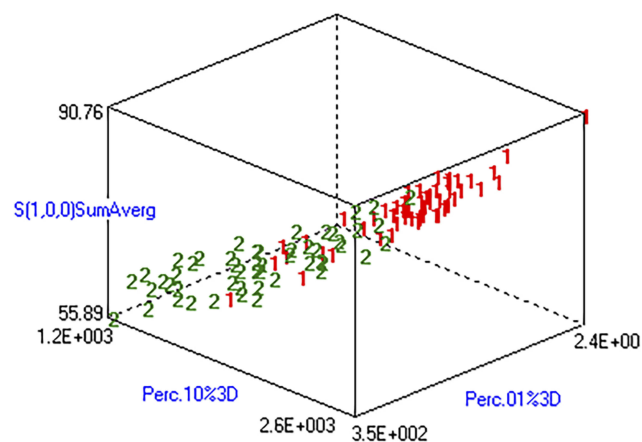
histological subtypes may be beneficial in preoperative evaluation of tumor aggressiveness and in determining therapeutic strategies.

**Table 2** Diagnostic Performance of Clinical Features and CT Texture for Differentiating SRCC from AC

	Variables	AUC	Sensitivity	Specificity	Cutoff Value	95% CI
Clinical features	CA724	0.745	0.569	0.922	0.490	0.647–0.843
Texture features	Perc.01%3D	0.927	0.765	0.961	0.726	0.881–0.974
	Perc.10%3D	0.929	0.784	0.961	0.745	0.880–0.978
	s(1,0,0)SumAverg	0.892	0.843	0.882	0.725	0.825–0.959



**Figure 4** The ROC analysis of Perc.01%3D, Perc.10%3D and s(1,0,0) SumAverg for differentiation SRCC from AC.  
**Abbreviation:** ROC, receiver operating characteristic.



**Figure 5** Texture analysis based on portal venous phase image to distinguish SRCC from AC. 1 for AC, 2 for SRCC. The difference between "1" and "2" is obvious, indicating excellent texture analysis and discrimination ability.

Our data show that males dominated in the SRCC group, whereas the location of the tumor did not show significant differences between SRCC and AC. While previous studies reported that SRCC was more commonly found in the right-sided colon and had a male predominance.<sup>22</sup> In our study, s(1,0,0) SumAverg was lower in the SRCC group than in the AC group. The parameter s(1,0,0) SumAverg can be extracted from the cooccurrence matrix, which is used to reflect uniformity/heterogeneity of tissue density by describing varies density variation with increasing distance in an

ROI.<sup>21,23</sup> Our data demonstrate that SRCC seems to behave as a heterogeneous tumor compared with AC, which was in agreement with a previous report.<sup>24</sup> The results of our study indicate that low Perc.01%3D and Perc.10%3D parameters were more frequent in the SRCC group than in the AC group. The 1st and 10th percentiles are measures that contain 1% and 10% minimum of the CT values in the observed data, which may represent the portion of tiny tumor necrosis that cannot be recognized by conventional imaging in SRCC. This was consistent with a previous study reporting AC was more likely to have high-degree contrast enhancement than SRCC.<sup>25</sup> The degree of contrast enhancement was correlated with vascular angiogenesis, microvessel density and extracellular vascular penetrability.<sup>15,26,27</sup> We speculate that SRCC induces less neovascularity and blood perfusion, which may be related to an aggressive phenotype and resistance to both radiotherapy and chemotherapy. Medium texture analysis was obtained based on the low value of the 1-NN misclassified rate (MCR 29.41%) in the differentiation between SRCC and AC, indicating that CT texture analysis can be used as an auxiliary method to distinguish SRCC from AC.

At present, there is no fixed standard for the selection of scanning phase in CT image texture analysis. In our study, the texture characteristics were performed on portal phase images according to a previous report, as the portal



phase can better display the lesion scope and help to distinguish the relationship between the lesion and the mucosal surface.<sup>28</sup> In addition, a routine portal phase scan was required during monitoring and follow-up of CRC patients. In this study, the evaluation of SRCC and AC was performed on the entire tumor, as entire tumor analysis provides a more representative evaluation of tumor heterogeneity. Previous studies have compared two-dimensional vs three-dimensional measurements of single lesions. Some reports showing that three-dimensional is no better than two-dimensional measurements, as more layers of ROI will contain more invalid components.<sup>29</sup> While others report that the analysis of three-dimensional measurements contains the whole volume of the tumor and may reflect more information about tumor heterogeneity.<sup>30</sup>

To our knowledge, this study was the first to compare CT imaging texture features of patients between SRCC and AC. However, the limitations of this paper include the following. First, the sample size for the SRCC group is small, owing to the rarity of SRCC. Therefore, the training cohort and validation cohort were not involved in our study. Second, the texture characteristics were performed on portal phase images, as the portal phase can better display the lesion scope. Further studies that include a larger study population and multiphase CT imaging are necessary to confirm the findings of our study. Third, our study analyzed adenocarcinoma and signet-ring cell carcinoma, and mucinous adenocarcinoma was excluded due to mucus composition. Finally, only three-dimensional measurements were assessed, which may represent the heterogeneous characteristics of the whole tumor. The comparison between 2D and 3D measurements will be studied in the future.

## Conclusion

In conclusion, our data indicate that SRCC usually occurs more often in men than AC does. In addition, our data also show that texture parameters on portal phase images, including Perc.01%3D, Perc.10%3D and s(1,0,0) SumAverg, may be regarded as promising data to assess the differentiation between SRCC and AC.

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## Author Contributions

All authors contributed to data analysis, drafting or revising the article, gave final approval of the version to be published, and agree to be accountable for all aspects of the work. Yali Yue and Feixiang Hu are co-first authors.

## Disclosure

The authors report no conflicts of interest in this work.

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