

# Delta-radiomics and response to neoadjuvant treatment in locally advanced gastric cancer—a multicenter study of GIRCG (Italian Research Group for Gastric Cancer)

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**Background:** To predict response to neoadjuvant chemotherapy (NAC) of gastric cancer (GC), prior to surgery, would be pivotal to customize patient treatment. The aim of this study is to investigate the reliability of computed tomography (CT) texture analysis (TA) in predicting the histo-pathological response to NAC in patients with resectable locally advanced gastric cancer (AGC).

**Methods:** Seventy (40 male, mean age 63.3 years) patients with resectable locally AGC, treated with NAC and radical surgery, were included in this retrospective study from 5 centers of the Italian Research Group for Gastric Cancer (GIRCG). Population was divided into two groups: 29 patients from one center (internal cohort for model development and internal validation) and 41 from other four centers (external cohort for independent external validation). Gross tumor volume (GTV) was segmented on each pre- and post-NAC multidetector CT (MDCT) image by using a dedicated software (RayStation), and 14 TA parameters were then extrapolated. Correlation between TA parameters and complete pathological response (tumor regression grade, TRG1), was initially investigated for the internal cohort. The univariate significant variables were tested on the external cohort and multivariate logistic analysis was performed.

**Results:** In multivariate logistic regression the only significant TA variable was delta gray-level cooccurrence matrix (GLCM) contrast (P=0.001, Nagelkerke R<sup>2</sup>: 0.546 for the internal cohort and P=0.014, Nagelkerke R<sup>2</sup>: 0.435 for the external cohort). Receiver operating characteristic (ROC) curves, generated from the logistic regression of all the patients, showed an area under the curve (AUC) of 0.763.

**Conclusions:** Post-NAC GLCM contrast and dissimilarity and delta GLCM contrast TA parameters seem to be reliable for identifying patients with locally AGC responder to NAC.

Keywords: Stomach neoplasms; neoadjuvant therapy; multidetector computed tomography (MDCT)

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# Introduction

Although the overall incidence of gastric cancer (GC) has progressively decreased during the last decades, it still remains the fourth most common malignancy and the thirdleading cause of cancer-related death (1,2); nevertheless, in Western countries, the disease is often at an advanced stage at the time of diagnosis and in more than two thirds of cases is unresectable or metastatic (3-5). To date surgery remains the only potentially curative modality to treat GC; however, it has been observed that a number of patients with locally advanced gastric cancer (AGC), who received R0 gastrectomy with extended (D2) lymphadenectomy, still develop distant metastases and loco-regional recurrences (6). For this reason, neoadjuvant chemotherapy (NAC) has been introduced successfully in the treatment of locally AGC (7,8) and it is now considered a standard treatment of care in Europe, with its effect in decreasing the tumor size (down-sizing) and the tumor stage (down-staging), increasing the R0 resection rate (9) and improving both overall and progression-free survival (PFS) (10,11). Anyway, a large amount (40-64%) of patients submitted to NAC do not present response to therapy (12) and critical points are monitoring of response and early identification of responder patients to chemotherapy. Histological tumor regression grade (TRG) after chemotherapy is considered an important prognostic and objective parameter of response evaluation, however it can only be assessed after surgery (13,14). An early detection of non-responder patients is important to not delay curative surgery and to avoid the toxicity induced by chemotherapy which may lead to increased surgical complications (7,11).

Traditional computed tomography (CT) is the imaging modality usually employed for staging and restaging GC patients after NAC, thanks to its high spatial resolution, large availability and low-cost (15); however, to date, there is no gold-standard criteria to objectify response to treatment, since gastric lesion is hardly measurable (16,17) and therefore not correctly evaluated adopting RECIST (18-20).

An emerging tool, which may play an important role in this field, is the texture analysis (TA), which belongs to the field of radiomics. This technique, can be applied to conventional CT images and seems to be able to detect subtle differences in CT values which cannot be recognized by human eyes (21), providing quantitative data on tumor microenvironment by analyzing the distribution and relationship of pixel or voxel gray levels in the image (22-24).

The aim of this study was to test the value of TA, applied to CT images, in predicting histological response of locally AGC to NAC.

## Methods

#### Patient selection

This retrospective study was approved by the institutional review boards of our hospitals and written informed consent for medical photographs was obtained from all subjects. Abdominal CT examinations of 70 patients were retrospectively reviewed from a cohort of 121 patients treated with NAC followed by gastrectomy in 5 Italian Research Group for Gastric Cancer (GIRCG) centers, between June 2010 and February 2019.

The inclusion criteria were the following: (I) biopsyproven, locally AGC without distant metastases (i.e., clinical parameters  $T \ge 3$  and/or N+, M0); (II) NAC before surgery and radical resection; (III) an interval time between restaging CT and surgery  $\leq 30$  days; (IV) the availability of histopathological response (TRG) according to Becker et al. (25). Fifty-one patients were excluded because of an interval between restaging CT and surgery longer than 30 days (n=3), presence of metastatic disease at restaging CT (n=4), neoplastic involvement of the esophagus (n=2), clinical complications during chemotherapy requiring urgent surgery (n=3), inappropriate CT methodology or technical parameters (n=6), use of a CT scanner with a number of slice lower than 64 (n=25), impossibility to import CT images on software for image segmentation (n=4), unavailability of CT post-NAC (n=1), other technical problems such as corrupted digital supports (n=3). Finally, 70 patients [40 male; 30 female; mean age 62.9; standard deviation (SD) 10.5 years] were enrolled in this study. The population was divided in two groups as follows: 29 patients enrolled from Siena University Hospital were considered as internal cohort (group 1) and 41 from the other GIRCG centers (Forlì Hospital, Montevarchi Hospital, Brescia Hospital and Verona Hospital) as external cohort (group 2)

Population	Internal patients	External patients
Total patients	29	41
Gender	17 M, 12 F	24 M, 17F
Median age	63.3±11.5	62.7±9.8
Responders (Becker 1)	3	8
Non responders (Becker 2–3)	26	33
Lauren classification		
Intestinal	14	21
Diffuse	12	17
Mixed	2	2
Non classified	1	1

Table 1 Internal and external cohorts of patients

M, male; F, female.

Table 2 Different chemotherapy regimens and number of patients with percentage in relation to single cohort population

Chemotherapy regimen	Internal cohort	External cohort
ECF (epirubicin, cisplatin and 5-fluorouracil)	5 (17.2%)	6 (14.7%)
DOX (docetaxel, oxaliplatin and capecitabine)	15 (51.7%)	23 (56.1%)
EOX (capecitabine, oxaliplatin and ED epirubicin)	3 (10.3%)	4 (9.8%)
DCF (docetaxel, cisplatin and 5-fluorouracil)	1 (3.5%)	1 (2.4%)
2ECF + 4DCF	1 (3.5%)	-
CDDP (cisplatin + capecitabine)	-	1 (2.4%)
FOLFOX (folic acid, fluorouracil and oxaliplatin)	3 (10.3%)	5 (12.2%)
1DOX + 3DOF (docetaxel, oxaliplatin and fluorouracil)	-	1 (2.4%)
EOF (epirubicin, oxaliplatin and 5-fluorouracil)	1 (3.5%)	-

for independent external validation (*Table 1*). The different chemotherapy regimens used are reported in *Table 2*.

# CT and image analyses

All 70 patients underwent CT scans (CTs) within 1 week before the beginning of NAC and within a maximum of 30 days (mean  $20\pm6.4$  days) after completion of NAC before surgery. All patients who underwent the CTs had fasted for 8 hours. Stomach distention was obtained with air or with water; in particular, air distention was obtained by administering two pouches of effervescent granules per os, together with 10 mL of water, 3 minutes before the scan, while water distention was obtained by asking patients to drink 3 or 4 glasses of water (125 mL) immediately before the exam. All patients also received 1 mg of glucagon (GlucaGen, Novo Nordisk) or 20 mg of hyoscine butylbromide (Buscopan, Pharmamedix) intravenously, in order to induce gastric hypotonia. The same inflation technique was used both in staging and restaging CTs for each patient, in order to reduce the bias deriving from the different degree of gastric distention. CTs were acquired with a spiral technique by using a 64-detector row configuration (VCT, General Electric Healthcare, Milwaukee, USA in 29 cases and LightSpeed Plus, GE Healthcare, Milwaukee, USA in 41 cases). The following protocol was used: after a scout view, an unenhanced upper abdominal CT scan was acquired from the diaphragmatic domes to 2 cm below the lower

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CT technical parameters	Details		
Slice thickness (mm)	1.25 mm for 29 patients and 2.00 mm for 41 patients		
Beam pitch	0.9/1.3		
Reconstruction interval (IR)	At least half of the effective slice thickness		
Tube voltage (kVp)	120–140		
Reference mAs	200/250–500/600		

Table 3 CT technical parameters, with slice thickness referred to the late arterial phase

CT, computed tomography.

margin of the gastric body to confirm the distention of the stomach. Contrast-enhanced CTs were performed in the late arterial phase (start delay 45-50 s) in the upper abdomen and in the portal venous phase (start delay 70-80 s) from pelvic brim to thoracic inlet, after an intravenous injection of 2 mL/kg of nonionic contrast material (iodine concentration  $\geq 350$  mg/mL), followed by 40 mL of saline solution, using a semiautomated power injector (3.5-4.0 mL/s flow rate) with an 18/20-gauge needle in the antecubital vein. A delayed CT scan after 5 minutes was used to characterize uncertain liver lesions. An automatic current modulation tube was used to minimize radiation exposure. Images reconstruction was performed by using a standard reconstruction algorithm. CT technical parameters are reported in *Table 3*.

Images of post-contrast late arterial phase of both preand post-NAC CTs were retrieved from a picture archiving and communication system (PACS). Gross tumor volume (GTV) was manually contoured on each slice, in the axial plane, along the outer edge of gastric lesion, by using a dedicated software (RayStation), trying to exclude from segmentation vessels and gastric contents (Figure 1). Each volume was interpreted and segmented in consensus by two radiologists with 5- and 16-year experience in abdominal CT, respectively, blinded to the clinicopathological information such as TNM stage, degree of tumor differentiation and histopathological response according to Becker. The impact of variations on contouring was analyzed performing two delineation on a small sample set of patients (namely 20), and the TA parameters were tested for reliability with intraclass coefficient correlation (ICC) method both in the images before and after NAC. The delta TA parameters, meaning the difference between the same TA parameter extrapolated from images before and after NAC, for each TA parameter, were tested for reliability with ICC. Only reliable TA parameters (ICC >0.70, single

measure) were then selected.

All the analyzes for this work have been accomplished with LifeX Software<sup>®</sup> (25). TA parameters included features of gray-level co-occurrence matrix (GLCM), compacity, sphericity and indices from the gray-level histogram, for a total of 14 parameters.

Parameters were extracted by using the following constants: spatial resampling (X: 2 mm, Y: 2 mm, Z: 2 mm), intensity discretization (number of gray levels: 64, size of bins: 10), intensity rescaling (absolute, min bound: -1,000 HU, max bound: 3,000 HU).

#### Pathological evaluation

Post-operative histopathological findings were evaluated by expert gastrointestinal pathologists. At the macroscopic examination, tumors were classified according to the criteria proposed by Borrmann *et al.* (26).

At least five tissue blocks from the tumor site were taken if tumor was grossly visible; if the viable tumor was not grossly evident, the whole suspicious area was embedded with step sectioning at 5 mm (13).

Lymph nodes samples were distinguished in stations according to the Japanese Research Society for Gastric Cancer (JRSGC) classification (27).

TRG was evaluated according to the Becker classification as percentage of residual neoplasia in the macroscopically evaluated tumor bed. TRG1 was defined as complete or subtotal tumor regression with <10% residual tumor cells; TGR2 as partial tumor regression with 10–50% residual tumor cells; and TRG3 as minimal or absent regression with >50% residual tumor cells with or without signs of treatment effects (13,25). We considered patients with grade 1 as responders and patients with grade 2 or 3 as non-responders in this study, as there are several evidences showing that patients with complete or subtotal tumoral



**Figure 1** Example of gastric cancer contouring before neoadjuvant chemotherapy. GTV was manually drawn through a ROI on each CT slice (A,B,C,D) and not only where neoplasm was more evident. GTV, gross tumor volume; ROI, region of interest.

regression (TRG 1a and 1b respectively) have a better prognosis compared to those with partial or minimal tumor regression (TRG 2 and 3 respectively) (28,29).

Cancer staging and residual tumor in surgical margins (R) were classified according to American Joint Committee on Cancer (AJCC) 8th edition (30).

## Statistical analysis

The reliable TA parameters (ICC >0.70, single measure) were then correlated with the development of complete pathological response (TRG1), for the internal cohort of patients, by performing a univariate analysis (univariate logistic regression, with Bonferroni correction for the number of variables). We analyzed the correlation between the significant TA parameters and, if a correlation larger than 0.80 was observed (Pearson correlation), then the variable with the lowest univariable correlation with the endpoint was omitted, to avoid the risk of overfitting the

model and of multicollinearity (31) in the multivariate analysis (binary logistic regression), with a method similar to previous works (32).

Logistic regression analysis was optimized by using the internal cohort of patient, and the outcome of the testing data was then predicted within the optimized model also in the external cohort. The ROC curve was then extrapolated by the binary logistic regression of the two datasets (internal and external cohort). In multivariate analysis also tumor volume change before and after NAC was correlated with TRG1.

All the statistical analysis was conducted with SPSS software 23.0 and a P value <0.05 was considered statistically significant.

# **Results**

## Patient characteristics

Among the 70 patients enrolled, tumors were in the fundus

 Table 4 Texture features concerning pre- and post-neoadjuvant

 chemotherapy (NAC) and relative intraclass coefficient correlation

 (ICC) between the two readers

TA parameters	Pre-NAC	Post-NAC
Volume.ml	0.986*	0.977*
Volume.vx	0.982*	0.980*
Skewness	0.160	0.201
Kurtosis	0.206	0.244
Entropy	0.753*	0.873*
Energy	0.792*	0.925*
Sphericity	0.641	0.719*
Compacity	0.949*	0.938*
GLCM.homogeneity	0.793*	0.869*
GLCM.energy	0.897*	0.966*
GLCM.contrast	0.968*	0.886*
GLCM.correlation	0.989*	0.977*
GLCM.entropy	0.766*	0.867*
GLCM.dissimilarity	0.951*	0.868*

\*, significant reproducible values. TA, texture analysis; GLCM, gray-level co-occurrence matrix.

(n=5), body (n=23), antrum (n=28) cardia and body (n=8) or antrum and body (n=6) of the stomach. Histopathological analysis revealed the following GC histotypes, according to Lauren (33): 14 intestinal, 12 diffuse, 2 mixed and 1 non classified from the internal cohort and 21 intestinal, 17 diffuse, 2 mixed, and 1 non classified from the external cohort. Three out of 29 patients from the internal cohort and 8 out of 41 patients from the external cohort were classified as TRG1 according to Becker *et al.* (11 TRG1 in total). Six out of these 11 cases were intestinal, 4 diffuse and 1 non-classified GC.

#### TA and histopathologic correlation

The reliability analysis of TA parameters, performed with ICC, showed that 11 out of 14 parameters for pre-NAC images and 12 out of 14 TA parameters for the post-NAC images were significantly reproducible among the two different contourings (ICC >0.70, single measure) and were included in the further feature selection process (*Table 4*). Within the internal cohort, the following TA parameters were significantly correlated with the endpoint (TRG1):

post-GLCM contrast (P=0.017), post-GLCM dissimilarity (P=0.027), delta entropy (P=0.002), delta GLCM contrast (P<0.001), delta GLCM entropy (P=0.006) and delta GLCM dissimilarity (P<0.001) (*Figure 2*). Post-GLCM contrast and dissimilarity, as well as delta entropy and delta GLCM entropy were excluded because of a correlation greater than 0.8 each other; therefore, delta GLCM contrast and delta GLCM dissimilarity were considered for multivariate analysis and only delta GLCM contrast proved to be significant (P=0.001, Nagelkerke  $R^2$ : 0.546). All the significant variables on univariate analysis of the internal cohort, except delta GLCM entropy, proved to be significant in the external cohort (*Figure 3*). Multivariate logistic regression performed on the external dataset showed that delta GLCM contrast was the only significant

ROC curve was generated from the logistic regression of all the cohort of patients, showed an AUC of 0.763 (standard error: 0.098, P=0.006, lower bound: 0.571, upper bound: 0.954) (*Figure 4*).

parameter (P=0.014, Nagelkerke R<sup>2</sup>: 0.435) in the external

Tumor volume change was not significantly correlated with TRG1 (P=0.07).

## **Discussion**

cohort too.

The overall survival (OS) in AGC is still very poor, also after radical surgery and extended nodal dissection (34). For this reason, NAC has been successfully introduced in the treatment of locally AGC, in order to decrease tumor size and downstage the disease, increasing the R0 resection rate and improving both PFS and OS versus surgery alone (10,11). However, lack of response to NAC may delay curative surgery, and chemotherapy-induced toxicity may lead to increase surgical complications (12). Therefore, it would be of crucial importance to find a reliable way to distinguish responder from non-responder patients, so that non-responders may directly undergo to surgery after the early assessment through imaging during NAC.

Some works successfully investigated the feasibility of CT in predicting histopathological response of GC after NAC, by measuring tumor volume and maximum diameter reduction rate (35,36), despite visual assessment may be affected by a large inter and intraobserver variability and reduction of tumor volume is not always correlated with response to therapy. Also advanced imaging modalities, such as dual energy CT (DECT) and CT perfusion were investigated in predicting histopathological response of



**Figure 2** Boxplots showing the two post-neoadjuvant chemotherapy texture features significantly correlated to histopathologic outcome: GLCM contrast and GLCM dissimilarity. GLCM, gray-level co-occurrence matrix; TRG, tumor regression grade.



Figure 3 Boxplots showing the four delta texture features (relative percentage variations) significantly correlated with the histopathologic outcome: delta entropy, delta GLCM contrast, delta GLCM dissimilarity and delta GLCM entropy, the last significant only on the first group (internal cohort of patients). GLCM, gray-level co-occurrence matrix; TRG, tumor regression grade.

AGC after NAC (37-39), however, CT perfusion is still affected by some important issues (40,41), in particular regarding the reproducibility and the possibility to compare its results with different proprietary softwares (42,43) whereas DECT is affected by the fact that not all neoplastic lesions are hypervascular and that the reduction of vascularization after chemotherapy is not always strictly related to tumor response (44-46).

In this sense TA, analyzing the distribution of pixels or voxels gray level in digital images, makes possible to extrapolate mathematical parameters (texture features) which reflect the underlying structure of the objects



**Figure 4** ROC curve was generated from the binary logistic regression using only delta GLCM contrast as predictor variable. AUC resulted to be 0.763 (standard error: 0.098, P=0.006, lower bound: 0.571, upper bound: 0.954). ROC, receiver operating characteristic; GLCM, gray-level co-occurrence matrix.

showed in the image (21-23). The use of this technique in oncology could make images interpretation more objective, as it provides quantitative information concerning tumor heterogeneity, which are invisible to human eyes (22,23). This new technique has already been found to be a potential predictive biomarker of response to therapy or survival for several neoplasms such as lung (47), breast (48), kidney (49) and colorectal (50) cancer and also to predict treatment complications (51-55). Few studies have been conducted until now regarding TA application to GC; nevertheless, results have been encouraging. Yoon et al. demonstrated that higher values of contrast and variance and lower values of correlation and angular second moment (ASM) were associated with an increased tumor heterogeneity and better survival in patients with HER2-positive AGC who were treated with trastuzumab (56). Giganti et al. showed that pre-treatment TA can reflect tumor features at a cellular level and provides important information regarding the response rate to NAC (57). Moreover, further studies showed other potential roles of TA such as: differential diagnosis between histological subtypes of GC, evaluation of tumor differentiation degree, Lauren classification and vascular invasion (58,59).

The aim of this study was to assess the reliability of TA applied to CT in predicting histopathologic response to NAC in patients with locally AGC before surgery. No statistically significant correlation was found between any single pre-NAC TA parameter and histologic outcome, neither in the internal cohort, nor in the external one. Otherwise, results concerning both internal and external cohort of patients, showed that two post-NAC TA features (GLCM contrast and dissimilarity) were significantly correlated to the endpoint (P=0.017 and P=0.001 respectively and post GLCM dissimilarity P=0.027 and P=0.001 respectively), being tendentially higher in responder patients. Moreover, values of relative changes of TA features before and after NAC, and in particular delta GLCM contrast, were significantly different between responders and non-responder patients and tended to be higher in the first group; this finding seems to reflect an increase of the histopathologic heterogeneity of neoplastic tissue induced by treatment and confirm our preliminary results (60). According to our study, Rao et al. showed that the relative changes after chemotherapy of TA entropy and uniformity were significant predictors of histopathologic response to chemotherapy in patients with colorectal cancer liver metastases; interestingly, as well as in our case, the authors failed to demonstrate a benefit for pre-NAC TA parameters in predicting response (61).

In our study most results of the internal cohort were confirmed for the external cohort of patients, although CTs were performed with different setting; given that we didn't apply any filter to the images, we may suppose that TA features are not strictly influenced by CT technical parameters.

This study has some limitations. First of all, the number of patients was small, despite multicenter study; moreover, NAC schemes were different, even if similarly distributed in the two cohorts. Second, regarding data collection, in some cases the distention grade of the stomach was significantly different between pre- and post-NAC CTs and this factor may have caused under- or overestimation of GTV, especially when the lesion was very thin. Third, given that the stomach is a cave organ, it was sometimes hard to avoid the inclusion of gastric content or vessels in the contouring, operation that could have altered TA results. However, to our knowledge, this is the first study correlating variations of TA parameters, obtained from CT examinations, with histopathologic response (Becker's score) in patients with resectable AGC. Furthermore, to date, this is the only study in literature on TA in GC with an external validation cohort that allows to reach the target of reproducibility, hard to overcome in most TA studies due to different methodologies and software programs. Moreover, it is one of the few studies where TA was performed on the entire tumor volume instead of the largest two-dimensional

section. Surely, larger, prospective and multicentric studies, performed with the same standardized software, are needed to confirm our results.

In conclusion, the possible value of TA in the clinical practice of the treatment of GC is still to be clarified. This study suggests that TA is an imaging biomarker which can provide addictive information to conventional CT, since it reflects tumor heterogeneity variations that cannot be captured by human eyes. Post-NAC GLCM contrast and dissimilarity and delta GLCM contrast could potentially be predictive of response to NAC in patients with AGC. If these results would be further confirmed, this new technique may become a reliable tool to identify responder patients in a pre-surgical phase, suggesting stopping NAC and proceed to surgery in case of poor response.

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# Footnote

*Conflicts of Interest:* All authors have completed the ICMJE uniform disclosure form (available at http://dx.doi. org/10.21037/qims-20-683). The authors have no conflicts of interest to declare.

*Ethical Statement:* This retrospective study was approved by the institutional review boards of our hospitals and written informed consent for medical photographs was obtained from all subjects.

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