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Retrospective Study

Radiomics Signature: A Potential Biomarker for β -Arrestin1 Phosphorylation

Prediction in Hepatocellular Carcinoma

Che F et al. Radiomics Prediction of β -Arrestin1-Phosphorylation in HCC.

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Abstract

BACKGROUND

The phosphorylation status of β -arrestin1 influences its function as a signal strongly related to sorafenib resistance. This retrospective study aimed to develop and validate radiomics-based models for predicting β -arrestin1 phosphorylation in hepatocellular carcinoma (HCC) using whole-lesion radiomics and visual imaging features on preoperative contrast-enhanced computed tomography (CT) images.

AIM

To develop and validate radiomics-based models for predicting β -arrestin1 phosphorylation in HCC using radiomics with contrast-enhanced CT.

METHODS

Ninety-nine HCC patients (training cohort: n = 69; validation cohort: n = 30) receiving systemic sorafenib treatment after surgery were enrolled in this retrospective study. Three-dimensional whole-lesion regions of interest (ROIs) were manually delineated along the tumor margins on portal venous CT images. Radiomics features were generated and selected to build a radiomics score using logistic regression analysis. Imaging features were evaluated by two radiologists independently. All these features were combined to establish clinico-radiological (CR) and clinico-radiological-radiomics (CRR) models by using multivariable logistic regression analysis. The diagnostic performance and clinical usefulness of the models were measured by receiver operating characteristic and decision curves, and the area under the curve (AUC) was determined. Their association with prognosis was evaluated using the Kaplan–Meier method.

RESULTS

Four radiomics features were selected to construct the radiomics score. In the multivariate analysis, ALT level, tumor size and tumor margin on portal venous phase images were found to be significant independent factors for predicting β -arrestin1

phosphorylation-positive HCC and were included in the CR model. The CRR model integrating the radiomics score with clinico-radiological risk factors showed better discriminative performance (AUC = 0.898, 95%CI, 0.820 to 0.977) than the CR model (AUC = 0.794, 95%CI, 0.686 to 0. 901; P = 0.011), with increased clinical usefulness confirmed in both the training and validation cohorts using decision curve analysis. The risk of β -arrestin1 phosphorylation predicted by the CRR model was significantly associated with overall survival in the training and validation cohorts (log-rank test, P<0.05).

CONCLUSION

The radiomics signature is a reliable tool for evaluating β -arrestin1 phosphorylation which has prognostic significance for HCC patients, providing the potential to better identify patients who would benefit from sorafenib treatment.

Key Words: Hepatocellular carcinoma; Sorafenib resistance; β-Arrestin1 phosphorylation; Radiomics; Computed tomography; Overall survival

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Core Tip: The aim of this study was to develop and validate radiomics-based models for predicting β -arrestin1 phosphorylation in hepatocellular carcinoma (HCC). A total of 99 HCC patients (training cohort: n = 69; validation cohort: n = 30) were included, and the final clinico-radiological-radiomics model integrating the radiomics scores and clinico-radiological risk factors showed satisfactory discriminative performance (AUC = 0.898, 95%CI, 0.820 to 0.977). The preoperative prediction model can be used as a noninvasive and effective tool to help predict the outcome of HCC patients treated with sorafenib and identify patients who would benefit most from sorafenib treatment.

6 INTRODUCTION

Hepatocellular carcinoma (HCC) was the sixth most common cancer and the third leading cause of cancer-related death worldwide in 2020(1). Liver resection and transplantation are considered potentially curative methods for early-stage patients with well-preserved liver function. For advanced-stage HCC, systemic therapies such as multikinase inhibitors and immune checkpoint inhibitors, represented by sorafenib, have shown the potential to confer a survival advantage of 2–3 mo(2, 3). However, patients undergoing sorafenib treatment have a high resistance rate, which is still the greatest challenge and leads to a discouraging prognosis(3). Hence, identifying patients who are more likely to benefit from sorafenib treatment and discovering related biomarkers associated with sorafenib treatment response are urgently needed.

 β -Arrestins, including β -arrestin1 and β -arrestin2, are important regulators of seventransmembrane domain G-protein-coupled receptors (GPCRs). They can block subsequent G protein activation and result in receptor desensitization by phosphorylation/dephosphorylation of β-arrestin1 at a carboxyl-terminal serine, Ser-412(4, 5). The development of sorafenib resistance includes primary and secondary resistance, and the phosphorylation of ERK has been widely accepted to play an important role in both(6-8). Activation of AKT and epithelial-mesenchymal transition (EMT) also participates in acquired resistance and the signaling pathway mentioned above has a strong relationship with β -arrestin1(9-11). Wu *et al* revealed that β -arrestin1 enhances hepatocellular carcinogenesis by inflammation-mediated Akt signaling(12) and promotes HCC invasion and metastasis through p-ERK1/2 to mediate EMT(13). The phosphorylation status of β-arrestin1 influences its function in activating downstream receptors such as ERK1/2, forming a negative feedback loop(14). All these signals are highly related to sorafenib resistance(9, 15), which indicates that the expression of phosphorylated β -arrestin1 (p- β -arrestin1) may correlate with sorafenib resistance in HCC patients. Thus, the preoperative prediction of β -arrestin1

phosphorylation may help identify patients who could benefit from sorafenib treatment.

Radiomics is a newly emerging computational medical imaging method that allows for the quantitative analysis and translation of medical images(16, 17). Additionally, radiomics studies can provide insights into the depth and comprehensive characterization of tumor heterogeneity, with the underlying hypothesis that radiomics cab better characterize tumor heterogeneity(18-20). Preliminary studies have suggested that radiomics features can be useful for tumor lesion detection(18, 21) and are potentially predictive of the microenvironment and molecular status of tumors(16, 22, 23). Xu *et al* extracted radiomics signatures from contrast-enhanced computed tomography (CECT) images to build a risk model that showed good performance in microvascular invasion stratification and could well predict the clinical outcomes of HCC patients(24). To the best of our knowledge, the value of radiomics based on CECT images in predicting β-arrestin1 phosphorylation in HCC has not yet been reported.

The purpose of this study was therefore to develop and validate a radiomics-based model combining visual imaging and clinical features for the preoperative noninvasive prediction of β -arrestin1 phosphorylation and to further investigate its association with prognostic outcomes in HCC patients.

MATERIALS AND METHODS

Patients

This retrospective study was approved by the Institutional Review Board of West China Hospital and the requirement for informed consent was waived. Patients who had histologically proven HCC and received systemic treatment with sorafenib after surgery between January 2013 and April 2017 were retrospectively reviewed and consecutively recorded. The inclusion criteria were as follows: (1) age ≥18 years; (2) pathologically confirmed HCC; (3) interval between CECT imaging and surgery less than four weeks; (4) treatment naive [i.e., no hepatectomy, transcatheter arterial chemoembolization (TACE) or radiofrequency ablation (RFA) before CECT]; and (5) administration of

400 mg sorafenib twice a day after surgery with up to two dose reductions allowed (from 400 mg once daily to 400 mg every 2 days) for drug-related adverse events. The exclusion criteria were as follows: (1) incomplete or poor-quality CT images; (2) interrupted sorafenib treatment for longer than 48 h between the initiation of sorafenib and the first follow-up time point; and (3) death or loss to follow-up. Among the 146 eligible patients, 47 patients were excluded because CECT imaging was performed more than 4 wk before surgery (n = 13), the CT images were incomplete or of poor quality (n = 11), sorafenib treatment was interrupted for longer than 48 h between the initiation of sorafenib and the first follow-up time point (n = 8), or the patient was lost to follow-up (n = 15). Therefore, 99 patients were ultimately enrolled in this study. In addition, the investigated laboratory data within 7 days of the CT examination and clinical conditions were recorded, as shown in Figure 1.

In this study, consecutive patients who underwent surgery between January 2013 and March 2016 comprised the training cohort and were used to construct the nomograms, and patients who underwent surgery from April 2016 to April 2017 comprised the validation cohort.

Imaging techniques

CT imaging was performed by using multidetector CT scanners (Revolution, GE Healthcare, Milwaukee, USA; SOMATOM definition, Siemens Healthcare, Erlangen, Germany). Precontrast images were first obtained before contrast agent (iodine concentration, 300-370 mg/mL; volume, 1.5–2.0 mL/kg of body weight; contrast type, iopromide injection, Bayer Pharma AG) injection. Then, the arterial phase and portal venous phase were obtained with the following parameters: tube voltage, 100-120 kVp; tube current, 450 mA; slice thickness, 0.625 mm; pitch, 0.992:1; rotation speed: 0.5 s/rot; and ASIR-V: 30%. The arterial phase and portal venous phase were obtained at 25 s and 60 s after contrast injection.

Imaging evaluation

Three abdominal radiologists who were blinded to the histopathological results, clinical data, and survival outcomes reviewed all the CT images. The following imaging

features were assessed by these readers: 1) tumor margin, defined as a non-smooth margin with budding portion protruding into the liver parenchyma or infiltrative appearance at the tumor periphery, otherwise as smooth margin; 2) tumor size, defined as the maximum diameter, measured on arterial phase transverse images or portal venous phase images; 3) pseudocapsule, defined as a complete capsule with a uniform border around most or all of the tumor, unequivocally thicker or more conspicuous than the fibrotic tissue around background nodules, otherwise as incomplete integrity or not applicable; 4) multifocality; 5) arterial phase hyperenhancement; 6) portal venous/delay phase hypoenhancement; 7) radiologic evidence of necrosis; 8) radiologic evidence of cirrhosis; and 9) portal vein tumor thrombosis invasion. All examinations were performed using a workstation and recorded on a picture archiving and communication system.

Immunohistochemistry

Surgically resected specimens embedded in paraffin were cut into 4 µm-thick sections dewaxed, hydrated, and subjected to antigen retrieval. Subsequently, the tissue slides were incubated with primary antibodies using rabbit anti-human p- β -arrestin1 polyclonal antibody (Abcam Biotechnology, ab247229; diluted, 1:200) at 4°C overnight, followed by incubation with secondary antibodies (cat # K5007; Dako). Staining was performed with 3,3′-diaminobenzidine (DAB) and counterstained with hematoxylin. Two senior pathologists who were blinded to all radiological and clinical results independently selected five nonoverlapping and discontinuous regions to calculate the mean for statistical analysis. Variations in the results within a range of 5% were reassessed, and a consensus decision was made. With the threshold value of 5% (p- β -arrestin1 tumor cells/total tumor cells), cases with expression higher than 5% were considered p- β -arrestin1 positive.

Follow-up surveillance after surgical resection

The patients were consistently followed-up after liver resection at intervals of 3 to 6 mo based on α -fetoprotein (AFP) and imaging examinations, including ultrasound, CT or magnetic resonance imaging (MRI), and the time of disease-specific progression (local

recurrence or distant organ metastasis) and time of death were recorded. These survival data were collected by one radiologist using electronic medical records and follow-up imaging studies until June 31, 2020. Overall survival (OS) was measured as the interval from the date of surgery to the date of death from a disease-related cause or the latest follow-up. For patients who were alive at the latest follow-up, the data were censored.

Radiomics workflow

Regions of interest (ROIs) were manually delineated around the outline of the tumor slice by slice using ITK-SNAP software (version 3.6.0) and excluded necrosis and calcification in the tumors. Radiomics features were generated from the images using in-house scientific research 3D analysis software (Analysis Kit, version V3.0.0. R, GE healthcare). Two classes of feature extraction methods were extracted as follows: the original feature class and 14 filter classes (boxmean, additiveGaussiannoise, binomialblurimage, curvatureflow, boxsigmaimage, log, wavelet, normalize, laplaciansharpening, discreteGaussian, mean, specklenoise, recursiveGaussian and shotnoise). A total of 2600 features were extracted from the tumors. Two radiologists (readers 1 and 2) performed ROI segmentation in a blinded manner to assess interobserver reliability. Reader 1 repeated the feature extraction twice during a 1-week period to evaluate intraobserver reliability. The interobserver reliability and intraobserver reliability were assessed by obtaining the intraclass correlation coefficient (ICC). Features with ICC values >0.75 were selected for subsequent investigation. The feature selection process comprised the following three steps in the training group: variance analysis, Spearman correlation, and Lasso regression analysis. The radiomics score of each patient was calculated using this determined multivariable logistic regression model.

Prediction models of β -arrestin1 phosphorylation

For CT radiographic and clinical factors, predictors with P<0.05 in the univariate logistic analysis (P<0.05) were included. Multivariate logistic analysis, which was used to identify significant predictors based on a backward stepwise selection process with the Akaike information criterion, was employed to develop a clinical-radiological (CR)

model. In addition, a clinical-radiological-radiomics (CRR) model was constructed by multivariate logistic regression analysis, tests of the association with radiomics scores, clinical factor evaluations and CT imaging findings based on a backward stepwise selection process with the Akaike information criterion.

Statistical analysis

Categorical variables are summarized as frequencies and proportions, while continuous variables are expressed as the means and standard deviations or medians and interquartile ranges (IQRs). The differences in characteristics between groups were evaluated using Student's t test (normal distribution) and the Mann-Whitney U test (skewed distribution) for continuous variables and the chi-squared test or Fisher's exact test for categorical variables. OS curves were drawn by using the Kaplan-Meier method, and the difference in OS between groups was compared using the log-rank test. Inter-observer agreement was applied to assess the reliability of imaging analysis using the Kappa test; 0–0.2 represents slight, 0.21–0.40: fair, 0.41–0.60: moderate, 0.61–0.80: substantial, 0.81–1: excellent.

The discriminative performance of the prediction models was quantified by the area under the curve (AUC) of receiver operator characteristic (ROC) curves. Differences in the ROC curves were compared by using the DeLong test. Calibration curves were generated to assess the calibration of the prediction model with the Hosmer-Lemeshow test. The probabilities of net benefits were quantified by decision curve analysis (DCA) to evaluate the clinical application value of the prediction models.

The statistical analyses were implemented using R statistical software (version 3.4.2, http://www.R-project.org) and SPSS software (version 22.0, IBM), and two-sided P values < 0.05 were considered significant.

RESULTS

Patient characteristics

Of the 99 patients (male/female: 88/11; mean age, 51.53 ± 12.62 years, range 21 to 78 years) included in the study (training (n = 69) and validation (n = 30)), p- β -arrestin1 was

identified in 39 (39.4%) patients. The 3-year survival rates of p- β -arrestin1-positive and p- β -arrestin1-negative HCC patients were 38.5% and 31.7%, respectively. The Kaplan-Meier method showed that p- β -arrestin1-positive patients lived longer than p- β -arrestin1-negative patients (P<0.05 with the log-rank test). The clinical, pathological, and imaging characteristics of patients in the training and validation cohorts are summarized in Table 1.

Development of the radiomics score

Variance analysis using t test identified 15 radiomics features, assessed by Spearman rank correlation, and 4 features (boxsigmaimage_glrlm_RunLengthNonUniformity, wavelet_firstorder_wavelet-HLL-Skewness, wavelet_glcm_wavelet-HLH-Correlation, and wavelet_ngtdm_wavelet-LHL-Busyness) were chosen for logistic regression analysis (P>0.05). Variables with P<0.1 in the univariable logistic regression analysis were included in the multivariable regression model with backward stepwise selection using the Akaike information criterion. The radiomics score was calculated with the following formula: radiomics score = -0.3527+0.4748* boxsigmaimage_glrlm_RunLengthNonUniformity+0.7046*wavelet_firstorder_wavelet-HLL-Skewness-0.5697*wavelet_glcm_wavelet-HLH-Correlation+0.6471* wavelet_ngtdm_wavelet-LHL-Busyness.

Development of the predictive models

In total, 2 clinical characteristics (alanine aminotransferase (ALT) and aspartate aminotransferase (AST) levels), 2 imaging features (tumor size and tumor margin on portal venous phase images) and the radiomics score were identified by univariate analysis (all P < 0.1). In the multivariable logistic regression analysis, radiomics score (odds ratio (OR), 3.412; 95% confidence interval (CI), 1.562 to 7.453, P = 0.002), ALT level (OR, 0.159; 95%CI, 0.038 to 0.673, P < 0.012), tumor size (OR, 0.243; 95%CI, 0.059 to 1.003, P = 0.05) and tumor margin (OR, 0.170; 95%CI, 0.044 to 0.664, P = 0.011) significantly predicted β -arrestin1 phosphorylation (Table 2). Thus, the CR and CRR models were constructed by using the above aggressive features and the nomograms of the above multiparametric models are shown in Figure 2 A, B. Excellent interobserver

agreement was observed for the imaging feature evaluation, with Kappa values of 0.890 for tumor size and 0.789 for smooth tumor margin (Figure 3).

Predictive performance of the models

In the training cohort, the AUCs of the radiomics score, CR model and CRR model were 0.754 (95%CI, 0.640 to 0.868), 0.794 (95%CI, 0.686 to 0.901) and 0.898 (95%CI, 0.820 to 0.977), respectively. The CRR model had a significantly higher AUC than the radiomics score (P = 0.007) and the CR model (P = 0.011). In the validation cohort, the AUCs of the radiomics score, CR model and CRR model were 0.704 (95%CI, 0.454 to 0.953), 0.646 (95%CI, 0.411 to 0.880) and 0.735 (95%CI, 0.505 to 0.966), respectively. The diagnostic performance of the radiomics score and two models is shown in Table 3 and Figure 2 C, D. The calibration curve of all the models showed excellent agreement between the predictions and observations in both the training and validation cohorts (all P>0.05) (Figure 2 E, F). The decision curve showed that the CRR model had the largest overall net benefit compared with the treat-all-patients as p- β -arrestin1 positive and treat-none patients as p- β -arrestin1 negative across the full range of reasonable threshold probabilities (Figure 4).

Risk stratification with p- β -arrestin1 predicted by the CRR model

According to the risk of β -arrestin1 phosphorylation predicted by the CRR model, patients with p- β -arrestin1 positivity lived longer than those with p- β -arrestin1 negativity using the log-rank test (P<0.05) in both the training and validation cohorts (Figure 5).

DISCUSSION

In this retrospective study, a CT image-based model incorporating qualitative imaging features, clinical characteristics and quantitative radiomics features for predicting β -arrestin1 phosphorylation in HCC was generated. In addition, in patients treated with sorafenib, we found that p- β -arrestin1-positive HCC patients predicted by the CRR model were associated with better prognosis. The CRR model may serve as a

noninvasive and effective tool to predict HCC patients β -arrestin1 phosphorylation status and help select patients who are suitable for sorafenib treatment.

For predicting β -arrestin1 phosphorylation in HCC patients, radiomics features provided increased power (AUC = 0.754) and were indicated to be independent predictors for p- β -arrestin1 in the final CRR model (P = 0.005). Utilizing the radiomics method, the proposed CRR model vielded an improved diagnostic performance in the training cohort (AUC from 0.794 to 0.898) and validation cohort (AUC from 0.646 to 0.735), indicating that the combined radiomics approach may have greater value in postoperative β -arrestin phosphorylation prediction than clinico-radiological features. The reason why the CRR model achieved the best predictive performance can be explained by the fact that the final model includes both qualitative and quantitative imaging features to provide a comprehensive overview of the correlations of radiomics features with HCC pathological status and genomics characteristics(25). The radiomics signature included shape, intensity, and texture information, which can reflect the complexity of the properties of the target tissue. Previous studies have shown that imaging features, including texture features, are informative of the gene expression profiles of HCC lesions, which parallels the diversity of molecular activities (26). Hectors et al found that MRI radiomics features are highly associated with HCC immunooncological characteristics and can serve as noninvasive predictors of its status(27). A predictive nomogram incorporating a radiomics signature and other clinico-radiological factors showed significantly improved diagnostic performance cytokeratin19 stratification of HCC(28). However, to our knowledge, no other studies have investigated the possible value of quantitative analysis integrating clinical factors in predicting β-arrestin1 phosphorylation in HCC. Developed in the training cohort and applied to the validation cohort, the radiomics score based on CECT images combined with clinico-radiological factors could correctly identify the β -arrestin1 phosphorylation status of more than 86.7% of the patients in the training cohort and was well validated quantitative multiple-feature parameter for the β-arrestin1 to serve as phosphorylation-based risk stratification of HCC patients.Our study investigates the

predictive aspects of computational-assisted models for the preoperative prediction of β -arrestin1 phosphorylation status, which currently can now only be attained by invasive biopsy or surgery. This computational method can guide clinical management by identifying patients for targeted therapy, as most patients recommended for systematic treatment according to the Barcelona Clinic Liver Cancer algorithm are not candidates for surgery due to their poor condition(29).

The clinicopathologic features of preoperative serum ALT levels were significantly different in the p- β -arrestin1-positive and p- β -arrestin1-negative groups in this study. In line with the findings of previous studies, serum ALT is an important hepatic inflammation marker that is correlated with liver function. Hepatitis infection can simultaneously induce serum ALT increases and β -arrestin1 upregulation, and higher serum ALT levels are a feature commonly associated with this subtype of HCC(12). In clinical practice, serum ALT levels can be easily obtained and incorporated into a radiomics model for individualized risk estimation. A larger tumor size and nonsmooth tumor margins were also shown to be associated with p- β -arrestin1 expression. This finding is in accordance with previous studies showing that β -arrestin1 can promote hepatocellular proliferation via the Akt pathway, and HCCs with higher p- β -arrestin1 Levels are more likely to have an infiltrative growth pattern(13). Serum AST levels were associated with p- β -arrestin1 in the univariate analysis but not the multivariate models, probably because of a lack of statistical power due to the insufficient number of patients.

We also found that patients who were p- β -arrestin1-positive lived longer than those who were p- β -arrestin1-negative. Previous studies have revealed that high expression of β -arrestin1 contributes to tumor survival, proliferation, angiogenesis, invasion and metastasis and is associated with the prognosis of epithelial ovarian cancer, prostate cancer and lung cancer(30-34). Although the correlation of β -arrestin1 with HCC prognosis has not been investigated, β -arrestin1 has been shown to be positively related to HCC carcinogenesis and metastasis(12, 13). Evidence has shown that the sorafenib response is impaired in HCC with dysregulated phosphorylated ERK (p-ERK) and AKT

(p-AKT) activation and that suppression of ERK1/2 increases sorafenib sensitivity in several HCC cell lines(35-37), while β-arrestin1 can activate PI3K/Akt signaling by Akt phosphorylation and trigger ERK1/2 phosphorylation-mediated EMT in HCC. Moreover, hyperactive PI3K/AKT signaling has been reported to be one of the primary causes of EMT in HCC resistance to sorafenib(35, 36, 38). These studies indicate that PI3K/AKT signaling and p-ERK1/2-mediated EMT signal hyperactivity may function in β-arrestin1-induced HCC resistance to sorafenib and further influence the prognosis of HCC patients treated with sorafenib, which was consistent with a series of studies recently showing that β -arrestin1 expression had some correlation with resistance to therapy in several types of cancers, such as breast(39) ovarian(40, 41) and non-small-cell lung cancer (42). Increased phosphorylation of β -arrestin1 Leads to decreased levels of dephosphorylated β-arrestin1, which influences its function in the activation of downstream factors, such as p-ERK and p-AKT. Therefore, the phosphorylation status of β-arrestin1 has a critical role in HCC sorafenib resistance. Predicting p-β-arrestin1 can help to identify patients who are sensitive to this treatment and prevent unnecessary side effects.

There were some limitations in our study. First, this was a retrospective longitudinal cohort study and selection bias may exist due to the strict inclusion criteria. Although we performed internal validation, additional external validation is needed to facilitate the wider use of this predictive model. Second, our study was performed at a single institution, and the CT scanner in this study was not fixed in their protocol. However, this could be a strength in terms of the generalizability of the findings by reflecting actual clinical practice. Third, p- β -arrestin1 positivity was defined as a cutoff of 5% for tumor cells to avoid false-positive results. The association between our predictive model and the graded degree of p- β -arrestin1 immunopositivity should be further assessed.

CONCLUSION

In conclusion, CECT-based radiomics combining clinico-radiological factors achieved desirable results in the prediction of β -arrestin1 phosphorylation in HCC which showed prognostic value in patients treated with sorafenib. This finding suggests that CT radiomics may provide promising and noninvasive biomarkers for the evaluation of p- β -arrestin1 expression and may help identify the subset of HCC patients who are more sensitive to sorafenib treatment, thus potentially guiding personalized treatment strategies.

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