


A Practical Predictive Model Based on Ultrasound Imaging and Clinical Indices for Estimation of Response to Neoadjuvant Chemotherapy in Patients with Breast Cancer

Pingping Ye¹
Hongbo Duan¹
Zhenya Zhao²
Shibo Fang¹ 

¹Department of Ultrasonography, The Sixth Hospital of Ningbo City of Zhejiang Province, Ningbo, 315100, People's Republic of China; ²Department of Imaging, The First Hospital of Ningbo City of Zhejiang Province, Ningbo, 315010, People's Republic of China

Purpose: Clinical responses of neoadjuvant chemotherapy (NACT) are associated with prognosis in patients with breast cancer. The selection of suitable variables for the prediction of clinical responses remains controversial. Herein, we developed a predictive model based on ultrasound imaging and clinical indices to identify patients most likely to benefit from NACT.

Patients and Methods: We recruited a total of 225 consecutive patients who underwent NACT followed by surgery and axillary lymph node dissection at the Sixth Hospital of Ningbo City of Zhejiang Province between January 1, 2018, and March 31, 2021. All patients had been diagnosed with breast cancer following the clinical examination. First, we created a training cohort of patients who underwent NACT+surgery (N=180) to develop a nomogram. We then validated the performance of the nomogram in a validation cohort of patients who underwent NACT+ surgery (N=45). Multivariate logistic regression was then used to identify independent risk factors that were associated with the response to NACT; these were then incorporated into the nomogram.

Results: Multivariate logistic regression analysis identified several significant differences as to clinical responses of NACT, including neutrophil-lymphocyte ratio (NLR), body mass index (BMI), pulsatility index (PI), resistance index (RI), blood flow, Ki67, histological type, molecular subtyping, and tumor size. The performance of the nomogram score exhibited a robust C-index of 0.89 (95% confidence interval [CI]: 0.83 to 0.95) in the training cohort and a high C-index of 0.87 (95% CI: 0.81 to 0.93) in the validation cohort. Clinical impact curves showed that the nomogram had a good predictive ability.

Conclusion: We successfully established an accurate and optimized nomogram incorporated ultrasound imaging and clinical indices that could be used preoperatively to predict clinical responses of NACT. This model can be used to evaluate the risk of clinical responses to NACT and therefore facilitate the choice of personalized therapy.

Keywords: breast cancer, NACT, clinical response, nomogram prediction

Introduction

Globally, breast cancer is the most common malignant tumor among women and the second leading cause of cancer-related death every year.¹ The NACT is widely offered to breast cancer patients and is mainly used to reduce tumor burden and enable breast-conserving surgery.^{2,3} In the past decade, NACT has provided the opportunity to assess the response to treatment by an in vivo chemosensitivity test.⁴

Correspondence: Shibo Fang
Department of Ultrasonography, The Sixth Hospital of Ningbo City of Zhejiang Province, 1059 Zhongshan East Road, Ningbo, Zhejiang Province, People's Republic of China
Tel +86 0574-87996115
Email yepingping321@163.com

Indeed, several studies have demonstrated that breast cancer patients who achieve clinical responses after NACT exhibit improved prognosis.^{5–8} Nevertheless, only 30–73% of patients achieved clinical responses of NACT,⁹ meaning that the majority of patients do not benefit from NACT. Given these concerns, changes in the approach and patient selection that result in greater sensitivity have been recommended as necessary to support further treatment. Therefore, the development of a practical tool to predict the pathological response in patients with breast cancer after NACT is necessary.

Multiple models have been published predicting axillary pathological response after NACT in varying cohorts.^{10–23} Some baseline clinicopathological features and genotyping can predict the curative effect of NACT for breast cancer and further reflect the satisfied prognosis. However, the high costs associated with these genetic and molecular tests limit their utility in regular clinical practice.²¹ The economical and practical variables selected to establish the optimal prediction model could help select the patients who will benefit from NACT and recommend a tailored approach when choosing the initial treatment. Therefore, it is vital that more readily accessible pathologic, angiographic indices, or laboratory biomarkers associated with NACT treatment outcomes be identified.

In the present study, we established a nomogram that incorporated ultrasound imaging and clinical indices to predict pathological response to NACT. We then analyzed the predictive performance of this nomogram in a deviation cohort and then verified the performance in an internal validation cohort.

Patients and Methods

Patients Enrollment

Between January 1, 2018, and March 31, 2021, we prospectively collated data from consecutive patients who had been diagnosed with breast cancer at the Sixth Hospital of Ning Bo City of Zhe Jiang Province. All patients had received NACT before surgery. This study was approved by the Institutional Ethics Committee of the Sixth Hospital of Ning Bo City of Zhe Jiang Province, in compliance with the Declaration of Helsinki. Written informed consent was obtained from all participants before any treatment. All patients' information was anonymous. The inclusion criteria were as follows: (1) All patients were diagnosed with invasive breast cancer using hollow needle biopsy before chemotherapy; (2) patient had received preoperative ultrasound imaging and routine pelvic

examination, and (3) patient had undergone 2 to 6 cycles of NACT before surgery. We excluded: (1) patients who had a history of all other malignancies and any patient who had an incomplete set of medical data and (2) other types of neoadjuvant therapy, including radiotherapy and endocrine therapy. The flow chart of patient selection and data process is shown in Figure 1.

Evaluating the Safety and Efficacy of NACT

According to RECIST (version 1.1) criteria, the response to NACT is divided into four levels: (1) complete response (CR) in which the tumor completely disappeared; (2) partial response (PR) in which the diameter of the tumor was increased by at least 30%; (3) progressive disease (PD) in which the diameter of the tumor was increased by at least 20%; and (4) stable disease (SD); in which the diameter of the tumor did not shrink sufficiently to qualify for PR but did not increase sufficiently to qualify for PD.²⁴

Statistical Analysis

Continuous variables are expressed as mean (standard deviation) and compared using the two-tailed *t*-test or the Mann–Whitney test. Categorical variables were compared using the χ^2 test or Fisher's exact test. Univariate and multivariate logistic analyses were used to explore the risk factors for a pathological response. A nomogram was formulated based on results arising from the multivariate logistic regression analysis. The nomogram was based on the proportional conversion of each regression coefficient in the multivariate logistic regression to a 0 to 100-point scale. The effect of the variable with the highest β coefficient (the absolute value) was assigned 100 points. Points were added for all independent variables to create a total, which was then converted to predicted probabilities. Next, we used bootstrapping plots to calculate the concordance index (C-index) and area under the time-dependent receiver operating characteristic curve (time-dependent AUC) so that we could evaluate our ability to calibrate the curve. Typically, C-index and AUC values that exceeded 0.6 were suggestive of reasonable estimation. The cut-off point for risk stratifications was selected using X-tile. All analysis was performed using the Python programming language (version 3.9.2, Python Software Foundation, <https://www.python.org/>) and R Project for Statistical Computing (version 4.0.4, <http://www.r-project.org/>). All *P* values were two-tailed, and *P* < 0.05 was considered statistically significant.

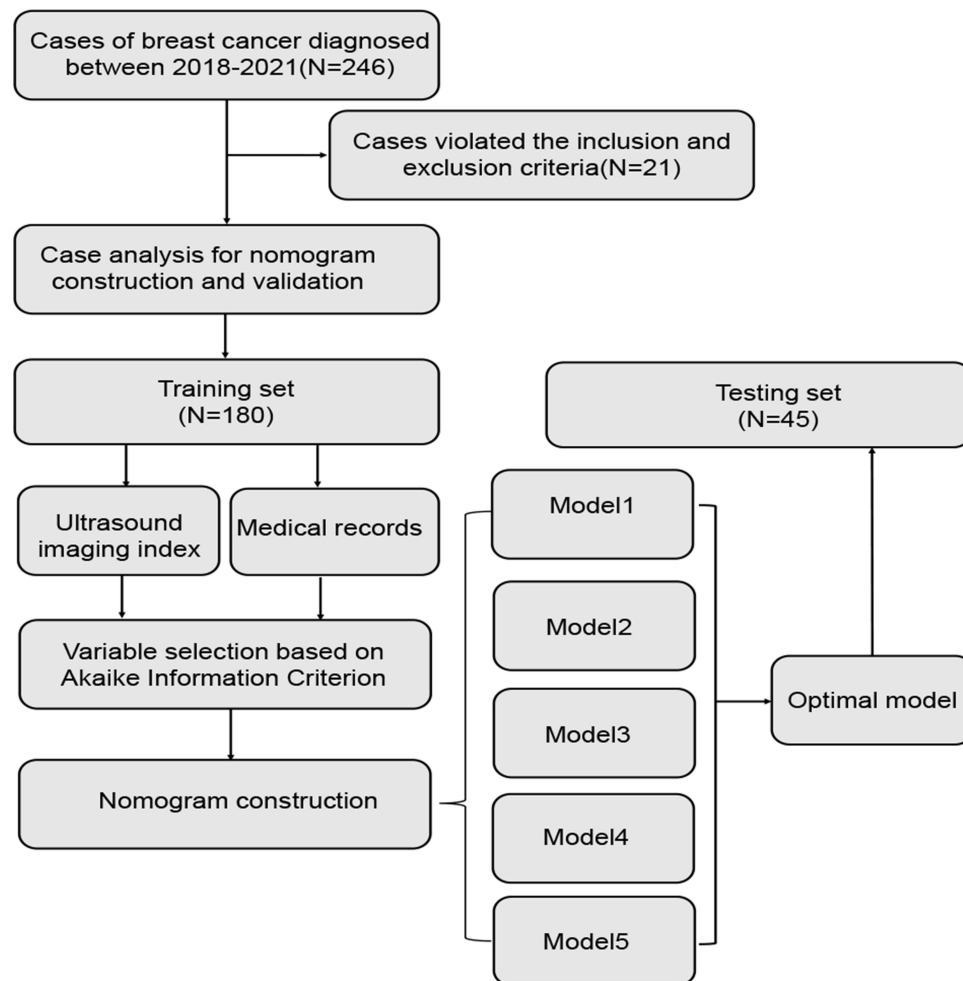


Figure 1 The flow chart of patient selection and data process.

Results

Clinicopathological Characteristics

During the period of enrollment, 246 consecutive patients with breast cancer underwent NACT in Sixth Hospital of Ning Bo City of Zhe Jiang Province. Of these, 180 patients met our inclusion criteria and were enrolled in our training cohort. Next, we enrolled in our validation cohort; these patients underwent procedures between May 1, 2020, and April 1, 2021 (n=45). Demographics and baseline data are summarized in Table 1. According to the RECIST (version 1.1) criteria, the clinical response was identified in 141 (78.3%) and 21 (82.2%) patients in the training and validation cohorts, respectively.

Feature Selection and Model Construction

Information criteria based on penalized likelihood, such as Akaike's information criterion (AIC), are widely used for model selection in health and biological research.²⁵ All the

parameters used to construct the nomogram were based on preoperative data, including peripheral blood inflammatory markers (eg, NLR), imaging indicators (eg, the largest tumor diameter, PI, RI, blood flow), and clinicopathological parameters (eg, BMI, histological type, molecular subtyping, Ki67, BMI). In the univariable analyses, variables with $P < 0.05$ were included in the multivariate logistic regression analysis. The results arising from the multivariate logistic analysis are summarized in Table 2. Based on the principle of AIC, we selected the top five AIC index models and summarized the intersection variables in the five models, including PI, RI, and blood flow (Figure 2A). Besides, the relative importance rank of all eight variables for estimation of pathological response and the five contributive models are illustrated in Figure 2B. The top-weighted features (BMI, NLR, tumor size, histological type, molecular subtyping, PI, RI, and blood flow classification) were highly correlated with poor pathological

Table I The Predictive Performances of Different Models Associated with Pathological Remission

Model	95% CI				Discrimination			
	OR	Lower	Upper	P-value	Brier	R ²	C-Index	AIC
Model1					0.04	0.06	0.882	325.70
Tumor_size*, cm	0.83	0.72	0.97	0.09				
PI <0.58;≥0.58	1.05	0.87	1.27	0.02				
RI ≥0.47;<0.47	1.19	0.93	1.51	0.04				
Pathology IDC:SC ILC:SC MCB:SC	0.30 0.48 1.52	0.14 0.35 0.93	0.63 0.66 2.49	0.01 0.02 <0.01				
Molecular_subtyping Luminal:TNBC HER2(+):TNBC LH:TNBC	0.95 0.58 0.74	0.74 0.45 0.55	1.21 0.75 0.99	<0.01 0.03 0.03				
Blood_flow ≥1.88;<1.88	1.34	1.02	1.78	0.04				
NLR ≥3.14;<3.14	1.18	0.67	2.09	<0.01				
BMI <18:18–24 ≥25:18–24	0.98 1.53	0.72 1.18	1.35 1.98	<0.01 0.02				
Model2					0.05	0.06	0.857	331.50
Tumor_size*, cm	0.78	0.39	1.17	0.02				
PI <0.58;≥0.58	1.23	0.84	1.62	0.01				
Ki67 ≥0.50;<0.50	1.14	0.75	1.53	0.04				
Pathology IDC:SC ILC:SC MCB:SC	0.48 0.52 1.39	0.03 0.11 0.94	0.93 0.93 1.84	0.04 <0.01 <0.01				
Molecular_subtyping Luminal:TNBC HER2(+):TNBC LH:TNBC	0.47 0.51 0.38	0.06 0.10 0.09	0.88 0.92 0.67	<0.01 0.02 0.04				
Blood_flow ≥1.88;<1.88	1.25	0.80	1.71	0.01				
NLR ≥3.14;<3.14	1.13	0.67	1.53	<0.01				

(Continued)

Table I (Continued).

Model	95% CI				Discrimination			
	OR	Lower	Upper	P-value	Brier	R ²	C-Index	AIC
BMI <18:18–24 ≥25:18–24	0.89 1.27	0.46 0.86	1.32 1.68	0.06 0.03				
Model3					0.04	0.06	0.764	334.60
Age*, year	0.67	0.16	1.18	0.05				
PI <0.58:≥0.58	0.85	0.42	1.28	0.02				
RI ≥0.47:<0.47	1.22	0.81	1.63	0.04				
Pathology IDC:SC ILC:SC MCB:SC	0.38 0.73 1.25	0.32 0.57 0.90	0.79 0.89 1.60	0.01 0.02 0.05				
Molecular_subtyping Luminal:TNBC HER2(+):TNBC LH:TNBC	0.79 0.58 0.24	0.46 0.29 0.11	1.12 0.87 0.59	0.02 0.03 0.01				
NLR ≥3.14:<3.14	1.09	0.60	1.58	<0.01				
Model4					0.05	0.07	0.775	332.20
Tumor_size*, cm	0.88	0.47	1.29	0.02				
Age*, year	1.07	0.66	1.48	0.05				
Ki67 ≥0.50:<0.50	1.22	0.75	1.69	0.03				
Pathology IDC:SC ILC:SC MCB:SC	0.46 0.45 1.21	0.30 0.31 0.79	0.62 0.59 1.62	<0.01 <0.01 <0.01				
Molecular_subtyping Luminal:TNBC HER2(+):TNBC LH:TNBC	0.85 0.57 0.77	0.48 0.02 0.36	1.22 1.12 1.18	0.02 <0.01 <0.01				
Blood_flow ≥1.88:<1.88	1.25	0.84	1.66	0.01				
NLR ≥3.14:<3.14	1.17	0.82	1.52	<0.01				
BMI <18:18–24 ≥25:18–24	0.91 1.03	0.53 0.64	1.28 1.42	<0.01 <0.01				

(Continued)

Table 1 (Continued).

Model	95% CI				Discrimination			
	OR	Lower	Upper	P-value	Brier	R ²	C-Index	AIC
Model5					0.04	0.05	0.796	334.50
Tumor_size*, cm	1.12	0.77	1.47	0.05				
Pathology								
IDC:SC	0.67	0.32	1.02	0.04				
ILC:SC	0.58	0.40	0.76	0.02				
MCB:SC	1.42	1.05	1.77	<0.01				
Molecular_subtyping								
Luminal:TNBC	0.78	0.37	1.19	<0.01				
HER2(+):TNBC	0.68	0.31	1.05	<0.01				
LH:TNBC	0.67	0.32	1.02	<0.01				
Ki67								
≥0.50:<0.50	1.24	0.81	1.67	0.02				
NLR								
≥3.14:<3.14	1.13	0.80	1.46	0.04				
BMI								
<18:18–24	0.72	0.41	1.03	<0.01				
≥25:18–24	0.98	0.49	1.47	<0.01				

Note: *Continuous variable. Clinicopathological characteristics of 225 breast cancer patients treated with neoadjuvant chemotherapy.

Abbreviations: AIC, Akaike Information Criterion; IDC, invasive ductal carcinoma; ILC, invasive lobular carcinoma; SC, simplex carcinoma; MCB, medullary carcinoma of breast; LH, luminal-HER2(+); TNBC, triple negative breast cancer; BMI, body mass index; PI, pulsatility index; RI, resistance index; NLR, neutrophil lymphocyte ratio; OR, odds ratio; 95% CI, 95% confidence interval.

response in breast cancer patients who received NACT. Multivariate logistic analysis based on AIC and candidate variables are summarized in [Table 1](#).

Development and Validation of a Predictive Nomogram for Pathological Response

Next, we optimized the model by performing a stepwise regression analysis. Multivariate analysis showed that tumor size (0.83; 95% CI, 0.72 to 0.97), PI (1.05, 95% CI, 0.87 to 1.27), RI (1.19, 95% CI, 0.93 to 1.51), pathology (IDC:SC, 0.30; 95% CI, 0.14 to 0.63; ILC:SC, 0.48; 95% CI, 0.35 to 0.66; MCB:SC, 1.52; 95% CI, 0.93 to 2.49), molecular subtyping (Luminal:TNBC, 0.95; 95% CI, 0.74 to 1.21; HER2(+):TNBC, 0.58; 95% CI, 0.45 to 0.75; LH:TNBC, 0.74; 95% CI, 0.55 to 0.99), blood flow (1.34; 95% CI, 1.02 to 1.78), NLR (1.18; 95% CI, 0.67 to 2.09), BMI (<18:18–24, 0.98; 95% CI, 0.72 to 1.35; ≥25:18–24, 1.53; 95% CI, 1.18 to 1.98) were independently associated with pathological response.

Assessment and Validation of the Models

The performance of the nomogram score exhibited a robust C-index of 0.89 (95% confidence interval [CI]: 0.83 to 0.95) in the training cohort and a high C-index of 0.87 (95% CI: 0.81 to 0.93) in the validation cohort. Calibration curves clearly showed that the nomogram produced robust results in the training cohort and testing cohort, respectively ([Figure 3](#)). Besides, incorporating ultrasound imaging and clinical indices achieved an AUC of 0.89 (95% CI: 0.9464–0.9778) in the identification of pathological response with an accuracy of 0.89 (95% CI: 90.1–94.4%) depending on individual clinical indicators or with an accuracy of 0.89 (95% CI: 90.1–94.4%) via ultrasound imaging index ([Figure 4A](#)). To further test the efficacy of the model, we used clinical impact curves to evaluate stratified risk. The calibration curve showed that the risk of pathological response could be readily distinguished in patients with breast cancer ([Figure 4B](#)).

Discussion

Many studies to date have evaluated the predictive ability of markers as a means of assessing breast cancer patients

Table 2 The Predictive Ability and Parameter Inclusion of Prediction Models Reported in Previous Literature

Variables	OR	95% CI		P-value
		Lower	Upper	
Tumor_size* , cm	0.83	0.72	0.97	<0.01
PI <0.58:≥0.58	1.05	0.87	1.27	0.02
RI ≥0.47:<0.47	1.19	0.93	1.51	0.04
Pathology				
IDC:SC	0.30	0.14	0.63	0.01
ILC:SC	0.48	0.35	0.66	0.02
MCB:SC	1.52	0.93	2.49	<0.01
Molecular_subtyping				
Luminal:TNBC	0.95	0.74	1.21	<0.01
HER2(+):TNBC	0.58	0.45	0.75	0.03
LH:TNBC	0.74	0.55	0.99	0.03
Blood_flow ≥1.88:<1.88	1.34	1.02	1.78	0.04
NLR ≥3.14:<3.14	1.18	0.67	2.09	<0.01
BMI				
<18:18–24	0.98	0.72	1.35	<0.01
≥25:18–24	1.53	1.18	1.98	0.02

Note: *Continuous variable. Multivariate logistic regression analysis for risk factors associated with rNACT in patients with breast cancer.

Abbreviations: IDC, invasive ductal carcinoma; ILC, invasive lobular carcinoma; SC, simplex carcinoma; MCB, medullary carcinoma of breast; LH, luminal-HER2(+); TNBC, triple negative breast cancer; BMI, body mass index; PI, pulsatility index; RI, resistance index; NLR, neutrophil lymphocyte ratio.

who achieved NACT.^{26–28} In early breast cancer, NACT can achieve an ideal complete or partial clinical response, which can lead to a better frequency of breast conserving therapy than adjuvant chemotherapy. Indeed, NACT can make breast conserving surgery more feasible, and is more likely to eradicate micrometastasis than NACT given after operation.²⁹ However, most patients did not achieve the desired expected results, and the reported clinical response rate was between 6% and 45%.^{30,31} In addition, the clinical response status is usually determined only by the surgical specimens after NACT. Herein, appropriate markers of chemosensitivity may help select patients who most benefit from NACT. As is shown in Table 2, studies have also demonstrated that clinicopathological parameters, particularly genetic and molecular phenotypes can be used to predict NACT treatment outcomes. However, the expensive cost, low efficiency of predicting

outcomes, and non-universal detection methods are not conducive to the vigorous promotion of these prediction models. As such, identifying reliable predictive factors for NACT in breast cancer remains a challenge because these parameters are not simple, conventional, objective, and inexpensive laboratory indexes. Collectively, the high costs associated with new biomarkers limited their utility in regular clinical practice.

Our study also demonstrated several factors that were significantly associated with pathological response and could be used as reliable predictive factors, including tumor size, PI, RI, pathology, molecular subtyping, blood flow classification, and BMI. Thus, we developed and validated a novel nomogram model to predict the probability of pathological response to NACT; we named this pathological response to NACT score (pre-rNACT score). This predictive model is a visualized scale that is calculated by weighted variables, including molecular biomarkers, medical imaging tests, and histomorphological factors. If breast cancer patients required NACT, then the clinician could rely on this pre-rNACT score to assess the risk of pathological response; this may help the clinician to decide on further treatments during the follow-up period. To verify whether patients could benefit from pre-NACT scoring, we used receiver operating characteristic curves and clinical impact curves. Based on pretreatment clinical factors and simple laboratory indexes, we found that the optimal decision cutoff point was achieved when the risk of pathological response to the NACT threshold was set to 0.57; this represented a point at which there was significant discriminative power. To our knowledge, we successfully used ultrasound imaging indicators to integrate clinicopathological indicators to predict the efficacy of NACT, which demonstrated that a robust model could help breast cancer patients evaluate the risk of pathological response to NACT.

In our study, we found that NLR was highly associated with pathological response, which was consistent with the results of a prior study.^{32,33} Indeed, inflammation is a key driver of cancer development and progression and is closely linked to metastasis.³⁴ Neutrophils acted as an important component of the tumor-induced inflammatory response, whereas lymphocytes are most often related to anti-tumor immune responses such that NLR values can reflect the dynamic balance between inflammatory and anti-inflammatory responses in cancer patients.²¹ Besides, the tumor immunoediting theory highlights the fact that different stages of the inflammatory response can

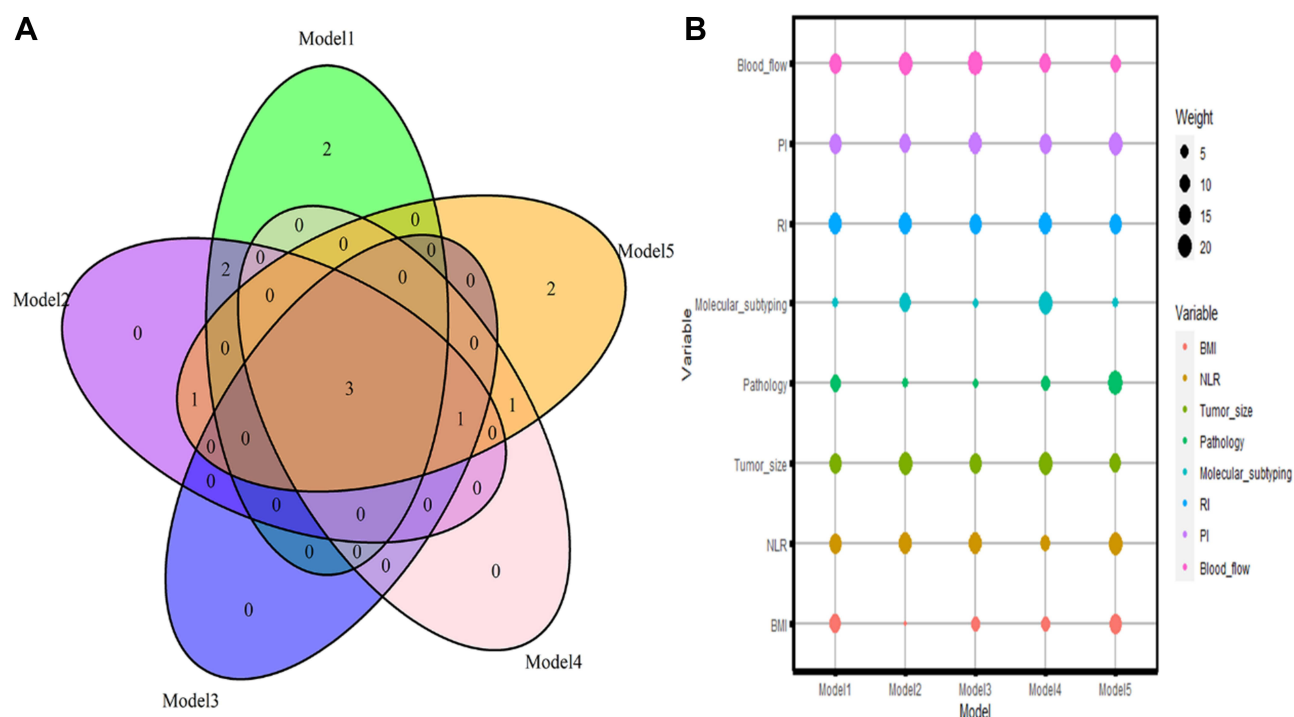


Figure 2 Statistical analysis of features included in models by Akaike Information Criterion. **(A)** Venn diagrams showing candidate variables for predicting the degree of pathological remission in five models. **(B)** Scaled importance rank of all features included in nomogram for identifying the possibility of pathological remission.

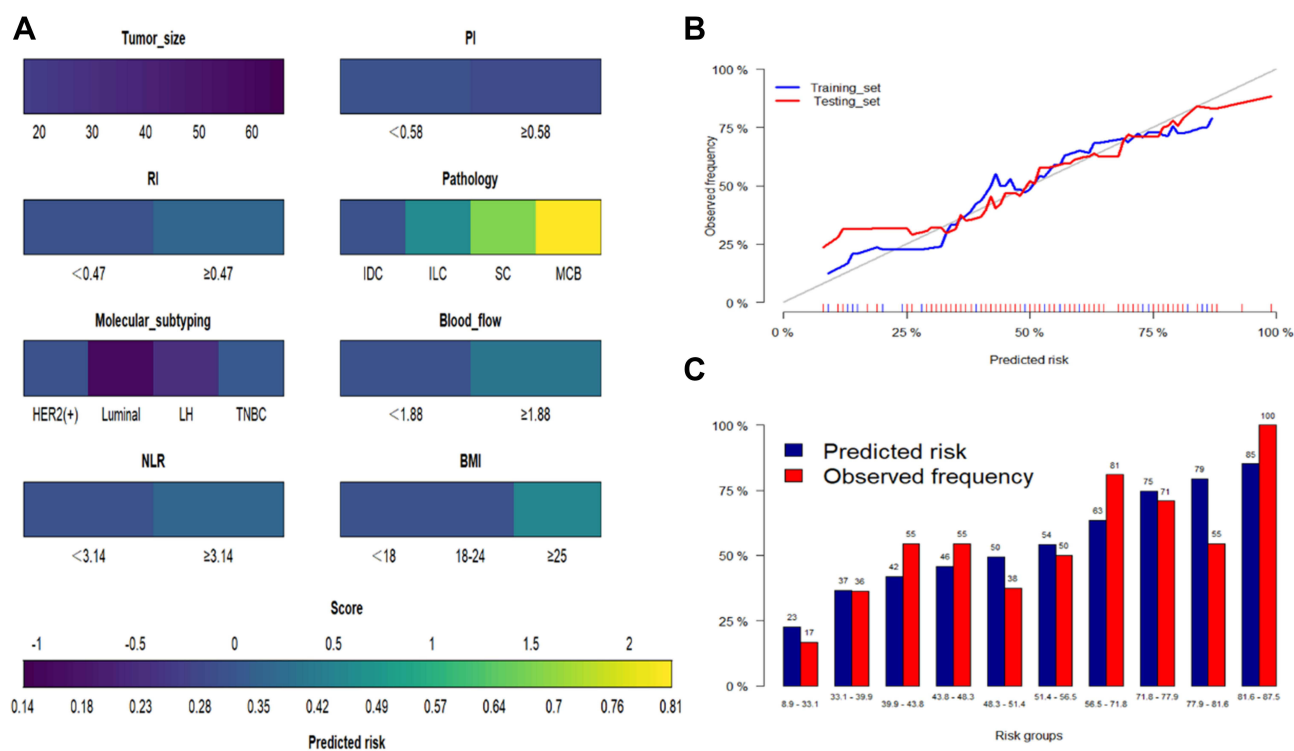


Figure 3 Generalized linear model. **(A)** Nomogram conveying the results of the candidate factors for predicting the possibility of pathological remission. **(B)** Calibration curves for internal validation of the nomogram (blue line represented training set, the red line represented testing set). **(C)** Predicted risk histogram comparing predicted risk of the nomogram with the observed frequency.

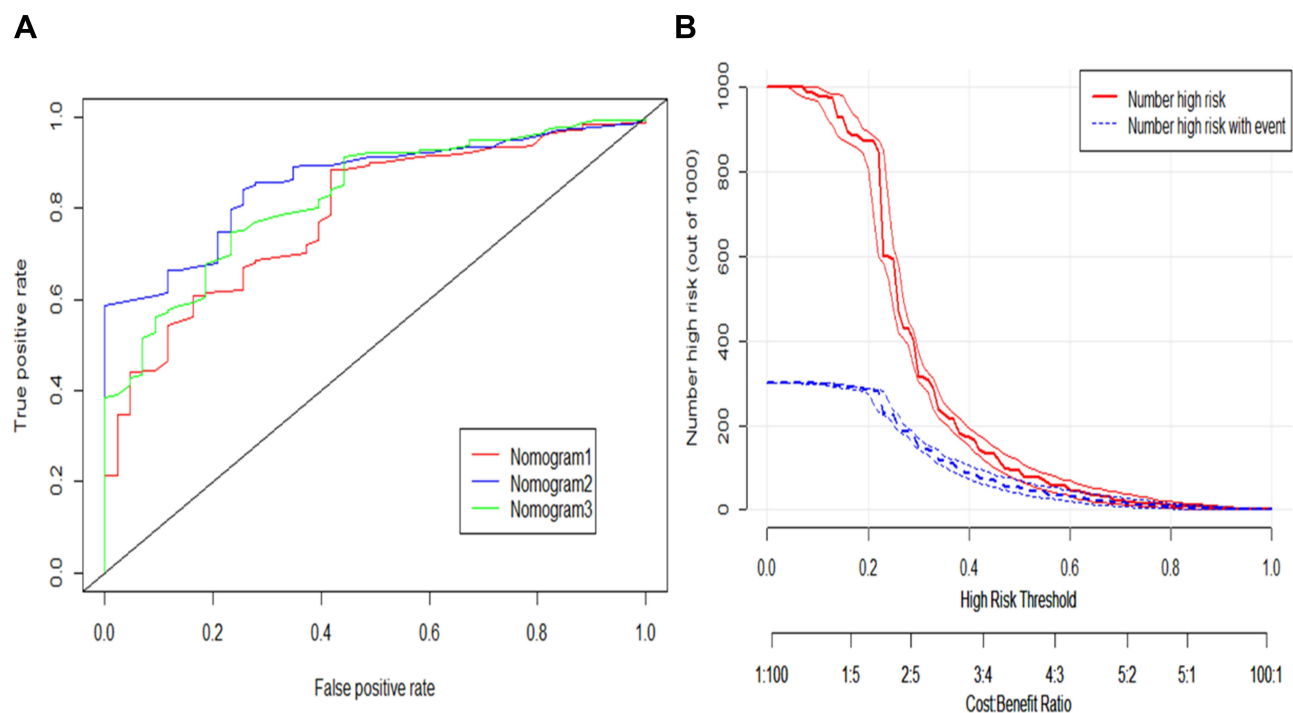


Figure 4 Predictive performance of nomogram based on ultrasound imaging and clinical indices. **(A)** AUC for predicting the possibility of pathological remission (Nomogram1. Ultrasound imaging index; Nomogram2. Ultrasound imaging index, Clinicopathological parameters; Nomogram3. Clinicopathological parameters). **(B)** Clinical impact curve for the nomogram score (nomogram2).

contribute to the generation and development of tumors.³⁵ In total, pretreatment NLR values served as a readout for inflammatory status, which has been employed to evaluate treatment efficacy and patient prognosis in cancer patients undergoing chemotherapy.

Previous studies have routinely used novel biomarkers to create nomograms. In this study, we sought to incorporate ultrasound image parameters, including PI, RI, and blood flow classification. Axillary ultrasound for restaging after NACT has been suggested to reduce the false-negative rates of sentinel lymph node biopsy.³⁶ However, even with the combined model constructed in previous studies through integration of imaging and clinicopathologic characteristics, the sensitivity and specificity for predicting axillary pathological response to NACT are moderate, 67.9%, and 73.6%, respectively.¹³ In contrast, we created an accurate nomogram by combining the results of clinicopathologic factors, imaging indicators, and serological tumor indicators. These variables are the most common clinical indicators and are therefore very easy to acquire. Other indicators can therefore be avoided, thus reducing the financial burden incurred by patients.

Nevertheless, the present study has several limitations that need to be considered. First, our observations were limited to retrospective studies from a single center, and it is now necessary to conduct external validation using data from other

centers. Second, the sample size was relatively small, and the predictive ability of the model needs to be further verified in large-sample studies. Third, because the model parameters were based on clinical peripheral blood and imaging results, the application of specific technologies (such as immunological diagnosis biomarkers) might improve the accuracy of our nomogram. Future research should be cautious and validated carefully.

Conclusion

We developed the pre-rNACT score scale, a novel nomogram-based prediction model featuring clinicopathologic factors and imaging results that can provide an optimized preoperative estimation of the risk of pathological response to NACT in patients with breast cancer.

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Disclosure

None of the authors have any conflicts of interest to declare.

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