



Original Research

Development and validation of two nomograms for predicting overall survival and cancer-specific survival in gastric cancer patients with liver metastases: A retrospective cohort study from SEER database

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ARTICLE INFO

Keywords:

Overall survival
Nomogram
Gastric cancer
Liver metastases

ABSTRACT

Background: Gastric cancer is heterogeneous and aggressive, especially with liver metastasis. This study aims to develop two nomograms to predict the overall survival (OS) and cancer-specific survival (CSS) of gastric cancer with liver metastasis (GCLM) patients.

Methods: From January 2000 to December 2018, a total of 1936 GCLM patients were selected from the Surveillance, Epidemiology, and End Results Program (SEER) database. They were further divided into a training cohort and a validation cohort, with the OS and CSS serving as the study's endpoints. The correlation analyses were used to determine the relationship between the variables. The univariate and multivariate Cox analyses were used to confirm the independent prognostic factors. To discriminate and calibrate the nomogram, calibration curves and the area under the time-dependent receiver operating characteristic curve (time-dependent AUC) were used. DCA curves were used to examine the accuracy and clinical benefits. The clinical utility of the nomogram and the AJCC Stage System was compared using net reclassification improvement (NRI) and integrated differentiation improvement (IDI) (IDI). Finally, the nomogram and the AJCC Stage System risk stratifications were compared.

Results: There was no collinearity among the variables that were screened. The results of multivariate Cox regression analysis showed that six variables (bone metastasis, lung metastasis, surgery, chemotherapy, grade, age) and five variables (lung metastasis, surgery, chemotherapy, grade, N stage) were identified to establish the nomogram for OS and CSS, respectively. The calibration curves, time-dependent AUC curves, and DCA revealed that both nomograms had pleasant predictive power. Furthermore, NRI and IDI confirmed that the nomogram outperformed the AJCC Stage System.

Conclusion: Both nomograms had satisfactory accuracy and were validated to assist clinicians in evaluating the prognosis of GCLM patients.

Introduction

Gastric cancer (GC) is a common clinical malignant tumor of the digestive tract that is responsible for the fourth and fifth leading causes of cancer-related deaths in men and women, respectively [1]. In 2020, more than one million (1,089,103) new cases of gastric cancer were

diagnosed worldwide, with an estimated 760,000 deaths [1]. There are several non-surgical therapies available today, including chemotherapy, radiotherapy, and immunotherapy. Furthermore, more researchers are focusing on tumor immunology and tumor-associated immune cells [2]. The use of a combination of Trastuzumab and chemotherapy in gastric adenocarcinoma patients with overexpressed human epidermal growth

Abbreviations: AUC, Area under the curve; CSS, Cancer-specific survival; DCA, Decision curve analysis; GC, Gastric cancer; GCLM, Gastric cancer and liver metastasis; IDI, Integrated differentiation improvement; LR, Liver resection; NRI, Net reclassification improvement; OS, Overall survival; ROC, Receiver operating characteristic; TG, Total gastrectomy; SEER, Surveillance, Epidemiology, and End Results Program.

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<https://doi.org/10.1016/j.tranon.2022.101480>

Received 5 May 2022; Received in revised form 4 June 2022; Accepted 4 July 2022

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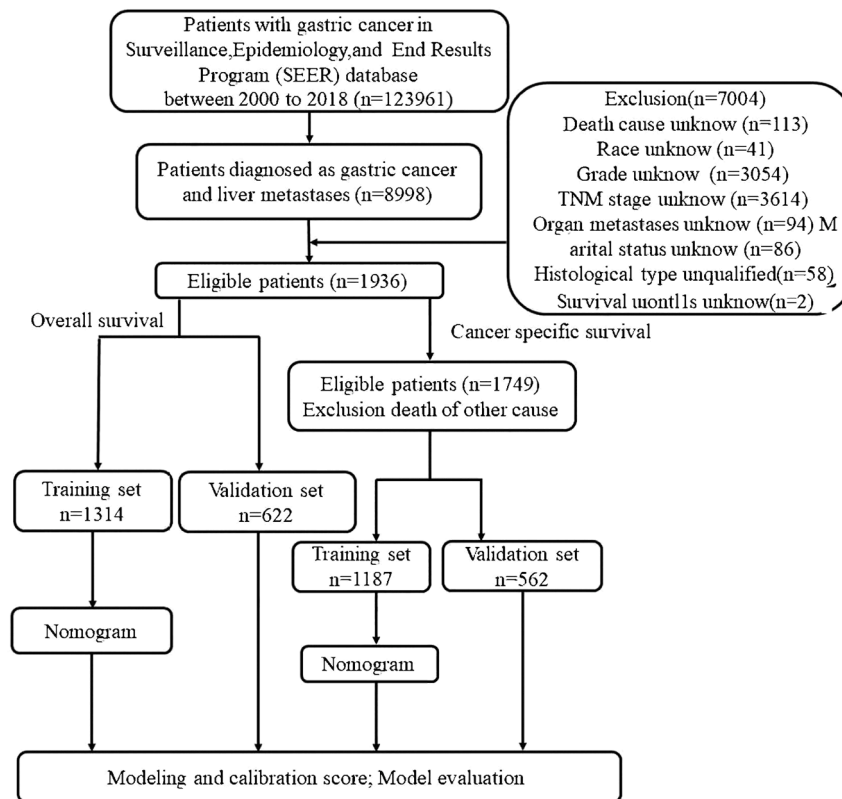


Fig. 1. Flow diagram illustrating recruitment of patients.

factor receptor 2 (HER2) has been a huge success and is considered a first-line treatment currently [3]. Moreover, his approach has made a significant improvement even as third-line therapy in advanced gastric cancer patients [4]. However, radical surgical resection is the primary treatment for localized GC [5]. Despite the development in therapeutic technology, recurrence rates in advanced cases remain high (40–80%) [6]. According to reports, approximately 35–40% of gastric cancer patients developed synchronous metastasis, with the vast majority of patients presenting with hepatic metastasis. Furthermore, after performing a curative surgical resection, more than 30% of GC patients developed metachronous liver metastasis [7,8]. With a 5-year survival rate of only 10%, liver metastasis from gastric cancer (GCLM) is an indicator of poor prognosis [9]. Currently, a lack of effective treatment modalities can improve overall survival [10].

The prognosis of GCLM varies considerably such that personalized prediction of GCLM has become the focus of various studies, including those of the American Joint Committee on Cancer (AJCC) gastric cancer staging system, which has confirmed the importance and practicability for the evaluation of the prognosis [11]. However, it is difficult to obtain satisfactory prediction outcomes with the TNM staging. More predictors and classification of continuous variables should be considered to improve the accuracy of prognosis.

The nomogram has been demonstrated to enhance predictive accuracy and widely used in oncology in recent years [12,13]. The nomogram model is simple, intuitive, and practical for visualizing the linear prognosis and quantifying individual patient survival in order to guide clinical decision-making and emphasize personalized medicine [14]. In this study, we aimed to develop a more detailed nomogram to predict the prognosis of GCLM patients using a large GCLM dataset from the Surveillance, Epidemiology, and End Results (SEER) database.

Materials and methods

Data source and inclusion criteria

The data for this study were abstracted using the SEER*Stat software version 8.3.9. The SEER database is a multi-center and multi-population registry funded by the National Cancer Institute that is not subject to medical ethics review and does not require informed consent. The data used for this study was extracted from the SEER Research Plus (with additional treatment fields) 18 Registry from 2000 to 2018, and it was subjected to strict inclusion and exclusion criteria, which are listed below. The inclusion criteria: (1) patients diagnosed with gastric cancers (Site recode of ICD-O-3/WHO2008:C160-C169); (2) liver metastases (SEER Combined Mets at DX-liver); (3) basic demographic variables, including age, race and gender; (4) complete survival data and follow-up data; (5) tumor characteristics, including histological information and type, TNM stage; and (6) therapeutic measures that whether received surgery, chemotherapy and radiotherapy. Individuals with gastric cancer who lacked information or a pathological diagnosis were excluded. Fig. 1 contains additional information. Furthermore, this work satisfies all STROCSS criteria [15].

Cohort definition and clinicopathological factors

We divided the data set into a 7:3 training cohort and validation cohorts using the R package (CreateDataPartition). The training set was used to create the model, while the validation set was used to optimize the model parameters and perform the evaluation. For the following variables, fifteen clinicopathological factors were extracted from the SEER database: age (<65 and ≥65 years), sex (female and male), race (White, Black, Asian or Pacific islander and American Indian/Alaska Native), primary site (cardia, fundus, body, gastric antrum, lesser, greater, other), histologic type (adenocarcinoma, signet ring cell, special type), grade (grade I, grade II, grade III, and grade IV), T stage (T0, T1,

Table 1
Demographic and clinical characteristics of patients with GCLM in OS group.

Characteristics	All samples N Percentage (%)	Training N Percentage (%)	Validation N Percentage (%)	P	
Age(years)					
<65 years	855	44.16%	584	44.44%	0.7172
>65 years	1081	55.84%	730	55.56%	
Sex					
female	538	27.79%	379	28.84%	0.1324
male	1398	72.21%	935	71.16%	
Race					
American Indian/Alaska Native	16	0.83%	10	0.76%	0.6237
Asian/Pacific Islander	209	10.80%	143	10.88%	
Black	323	16.68%	216	16.44%	
White	1388	71.69%	945	71.92%	
Primary Site					
Cardia	802	41.43%	551	41.93%	0.4580
Pylorus	95	4.91%	62	4.72%	
Body	169	8.73%	118	8.98%	
Antrum	308	15.91%	208	15.83%	
Fundus	28	1.44%	19	1.45%	
Lesser curve	119	6.14%	80	6.09%	
Greater curve	67	3.46%	46	3.50%	
Other	348	17.98%	230	17.50%	
Histologic type					
Adenocarcinoma	1609	83.11%	1089	82.88%	0.6867
Signet ring cell	137	7.08%	95	7.23%	
Special type	190	9.81%	130	9.89%	
Grade					
I	57	2.94%	41	3.12%	0.4192
II	621	32.08%	416	31.66%	
III	1214	62.71%	834	63.47%	
IV	44	2.27%	23	1.75%	
T Stage					
T0	3	0.15%	2	0.15%	0.7041
T1	759	39.20%	520	39.57%	
T2	110	5.68%	71	5.40%	
T3	461	23.81%	311	23.67%	
T4	603	31.15%	410	31.20%	
N Stage					
N0	756	39.05%	516	39.27%	0.7374
N1	864	44.63%	583	44.37%	
N2	158	8.16%	108	8.22%	
N3	158	8.16%	107	8.14%	
Radiotherapy					
Yes	359	18.54%	235	17.88%	0.2781
No	1577	81.46%	1079	82.12%	
Chemotherapy					
Yes	1171	60.73%	803	59.61%	0.4131
No	765	39.27%	511	40.39%	
Surgery					
Yes	298	15.39%	199	15.14%	0.6604
No	1638	84.61%	1115	84.86%	
Bone Metastasis					
Yes	157	8.85%	110	8.37%	0.2490
No	1618	91.15%	1204	91.63%	
Brain Metastasis					
Yes	26	1.34%	14	1.07%	0.1231
No	1910	98.66%	1300	98.93%	
Lung Metastasis					
Yes	315	16.27%	209	15.91%	0.5271
No	1621	83.73%	1105	84.09%	
Marital Status					
Yes	1191	61.52%	820	62.40%	0.2441
No	745	38.48%	494	37.60%	

T2, T3, and T4), N stage (N0, N1, N2 and N3), bone metastasis (yes or no), brain metastasis (yes or no), lung metastasis (yes or no), radiotherapy (yes or no), chemotherapy (yes or no), surgery (yes or no), marital status (yes or no). The collected follow-up data included overall survival (OS) and cancer-specific survival (CSS), which were considered endpoint times. We performed univariate Cox regression analysis on all fifteen prognostic factors and obtained independent prognostic factors through multivariate Cox regression analysis based on univariate Cox regression analysis ($P < 0.05$).

Statistical analysis

R software was used for all statistical analyses (version 4.1.0). We established three models: the Cox-AJCC model, the multi-factor Cox model and the competitive risk model. First, simple data processing was performed, converting raw data into factors for subsequent analysis. Pearson correlation was used to assess the existence of correlation among the variables in the correlation analyses. After randomly dividing the data into training and validation cohorts in a 7:3 ratio, univariate

Table 2
Demographic and clinical characteristics of patients with GCLM in CSS group.

Characteristics	All samples N Percentage (%)	Training N Percentage (%)	Validation N Percentage (%)	P	
Age(years)					
<65	773 44.20%	523 44.06%	250 44.48%	0.8678	
>65	976 55.80%	664 55.94%	312 55.52%		
Sex					
female	476 27.22%	337 28.39%	139 24.73%	0.1085	
male	1273 72.78%	850 71.61%	423 75.27%		
Race					
American Indian/Alaska Native	16 0.91%	10 0.84%	6 1.07%	0.9427	
Asian/Pacific Islander	194 11.09%	134 11.29%	60 10.68%		
Black	281 16.07%	189 15.92%	92 16.37%		
White	1258 71.93%	854 71.95%	404 71.89%		
Primary Site					
Cardia	741 42.37%	507 42.71%	234 41.64%	0.9731	
Pylorus	88 5.03%	58 4.89%	30 5.34%		
Body	153 8.75%	107 9.01%	46 8.19%		
Antrum	265 15.15%	182 15.33%	83 14.77%		
Fundus	26 1.49%	17 1.43%	9 1.60%		
Lesser curve	104 5.95%	70 5.90%	34 6.05%		
Greater curve	60 3.43%	42 3.54%	18 3.20%		
Other	312 17.84%	204 17.19%	108 19.22%		
Histologic type					
Adenocarcinoma	1452 83.02%	983 82.81%	469 83.45%		0.9448
Signet ring cell	127 7.26%	87 7.33%	40 7.12%		
Special type	170 9.72%	117 9.86%	53 9.43%		
Grade					
I	43 2.46%	31 2.61%	12 2.14%	0.0936	
II	553 31.62%	374 31.51%	179 31.85%		
III	1113 63.64%	762 64.20%	351 62.46%		
IV	40 2.29%	20 1.68%	20 3.56%		
T Stage					
T0	2 0.11%	1 0.08%	1 0.18%	0.9488	
T1	688 39.34%	472 39.76%	216 38.43%		
T2	99 5.66%	65 5.48%	34 6.05%		
T3	418 23.90%	283 23.84%	135 24.02%		
T4	542 30.99%	366 30.83%	176 31.32%		
N Stage					
N0	674 38.54%	459 38.67%	215 38.26%	0.9483	
N1	788 45.05%	537 45.24%	251 44.66%		
N2	144 8.23%	97 8.17%	47 8.36%		
N3	143 8.18%	94 7.92%	49 8.72%		
Radiotherapy					
Yes	325 18.58%	215 18.11%	110 19.57%	0.4635	
No	1424 81.42%	972 81.89%	452 80.43%		
Chemotherapy					
Yes	1061 60.66%	727 61.25%	334 59.43%	0.4678	
No	688 39.34%	460 38.75%	228 40.57%		
Surgery					
Yes	240 13.72%	161 13.56%	79 14.06%	0.7795	
No	1509 86.28%	1026 86.44%	483 85.94%		
Bone Metastasis					
Yes	157 8.98%	103 8.68%	54 9.61%	0.5246	
No	1592 91.02%	1084 91.32%	508 90.39%		
Brain Metastasis					
Yes	24 1.37%	13 1.10%	11 1.96%	0.1478	
No	1725 98.63%	1174 98.90%	551 98.04%		
Lung Metastasis					
Yes	297 16.98%	199 16.76%	98 17.44%	0.7264	
No	1452 83.02%	988 83.24%	464 82.56%		
Marital Status					
Yes	1086 62.09%	746 62.85%	340 60.50%	0.3443	
No	663 37.91%	441 37.15%	222 39.50%		

Cox analysis was used to screen independent variables based on P values ($P < 0.1$). To compare the significant factors and estimate the hazard ratios (HRs) and 95% confidence intervals following the standard of $P < 0.05$, the multivariate Cox regression analysis was performed in variables that exhibit differences in univariate Cox regression analysis. After that, using Cox regression, which is based on the training cohort, to analyze the nomogram that was created to predict the 1-year, 3-year, and 4-year OS and CSS rates. With the establishment of these models, we use the net reclassification index (NRI) and integrated discrimination

improvement (IDI) methods to evaluate the clinical benefits and utility of the Cox-AJCC model and the multivariate Cox model, to select the best predictive model. There are two mutually complementary validation methods, but the NRI, which is primarily used to compare the prediction ability of the old model with the new one, only considers the improvement when a specific cutoff point is set, whereas the IDI, which is primarily used to investigate the overall improvement of the model, inspects the overall improved performance of the model [16]. In the case of CSS, the Fine-Gray proportional hazards model was used to create the

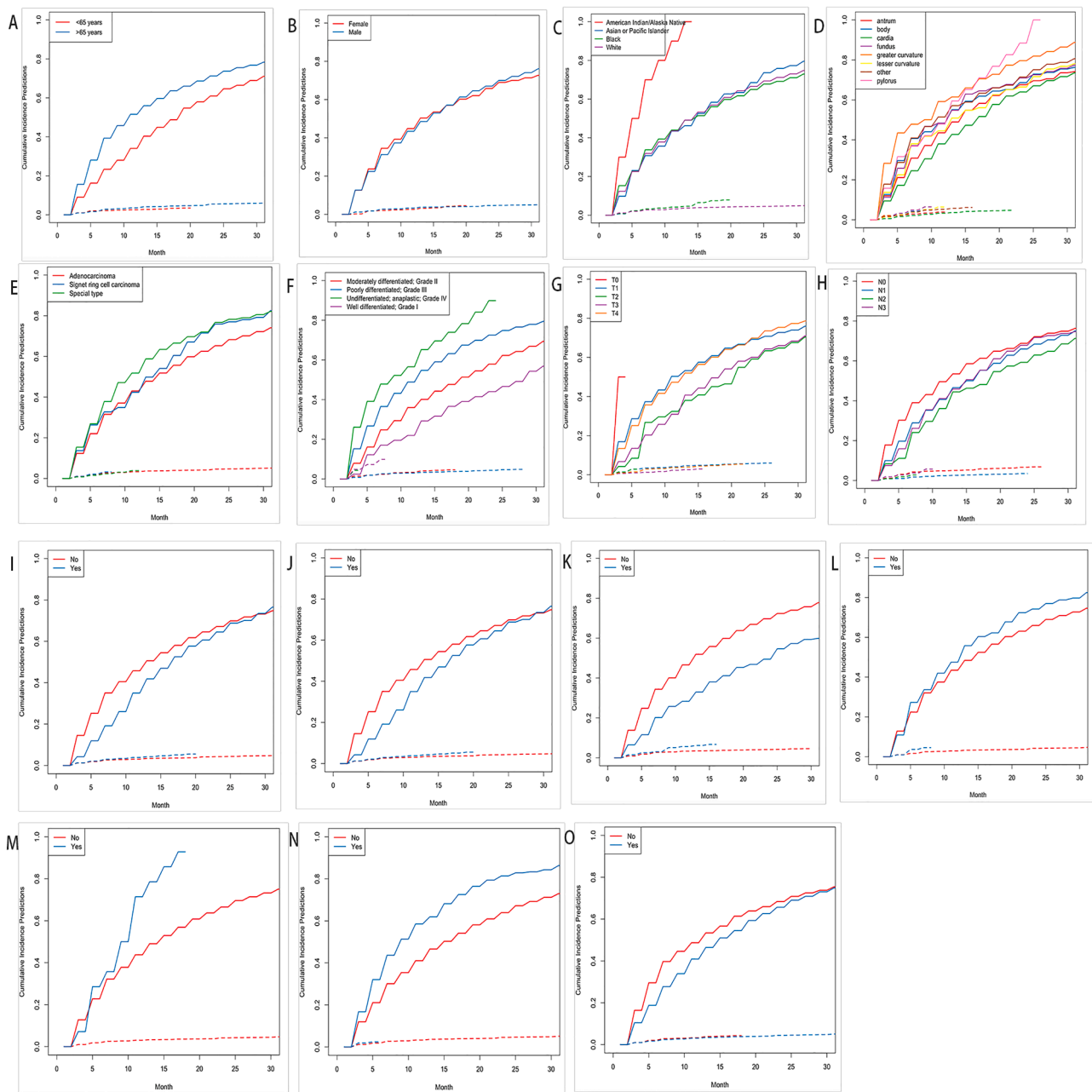


Fig. 2. Cumulative incidence predictions of CSS in gastric cancer with liver metastasis. (A) Age (B) Sex (C) Race (D) Primary Site (E) Histological type (F) Grade (G) T Stage (H) N Stage (I) Radiotherapy (J) Chemotherapy (K) Surgery (L) Bone Metastasis (M) Brain Metastasis (N) Lung Metastasis (O) Marital Status.

competing risk nomogram.

The discriminative power of the nomograms was evaluated using the area under the curve (AUC) values, which reflect the overall estimation value for all thresholds [17]. Lastly, the performance of our nomograms was investigated in the test and validation cohort in terms of the Calibration Curve, Receiver operating characteristic analysis (ROC), and Decision curve analysis (DCA).

Results

Flowchart

The flow diagram is displayed in Fig. 1.

Demographic and clinical features

There were 1936 patients in the study for OS analysis, with 1398 men and 538 women in this study. In addition, the patients were randomly assigned to one of two cohorts: training ($n = 1314$) and validation ($n = 522$). We described the demographic and clinical characteristics of GCLM patients. When first diagnosed, the majority of GCLM patients (55.84%) are poorly differentiated (Grade III) and over 65 years old. T1 (39.20%), T4 (31.15%), T3 (23.81%), T2 (5.68%), and T0 were the classifications (0.15%). More than half of the patients were male (72.21%) and white (71.69%), and the majority (84.61%) did not have surgery or radiotherapy (81.46%). Adenocarcinoma was diagnosed in 83.11% of the patients, and chemotherapy was used as their treatment (60.73%). Table 1 lists the demographic and clinical characteristics of patients in the OS group in detail.

A total of 1749 patients were included in the CSS analysis, with 1187

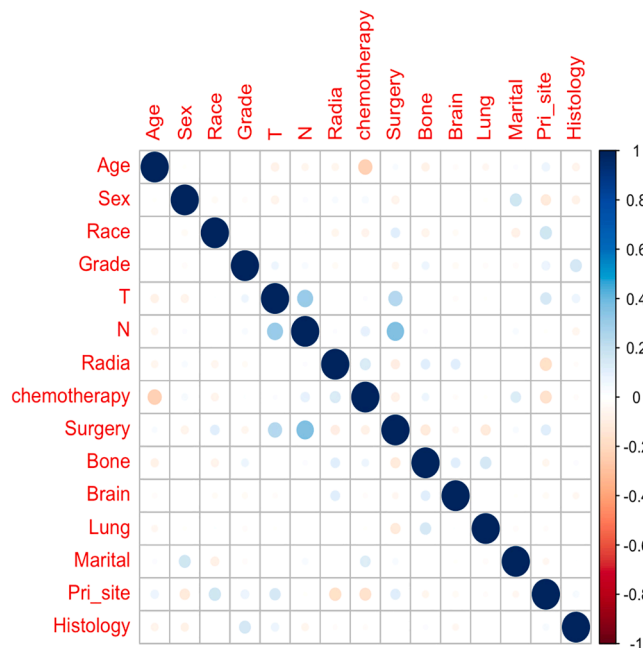


Fig. 3. The results of correlation analysis between all included variables.

patients in the training cohort and the remaining 562 patients in the validation cohort. There were 72.78% male patients and 27.22% female patients among these 1749 patients. The majority of the patients (71.93%) were white. 1086 patients (62.09%) were married, and less than 15% of patients underwent surgery as part of their treatment. Table 2 shows the baseline clinical-pathological characteristics of patients in the CSS group. The Cumulative Incidence Function (CIF) subgroup assessment data reported that high CSS occurred primarily in GCLM patients aged 65 years (Fig. 2A), American Indian/Alaska native (Fig. 2C), advanced grade (Fig. 2F), along with brain metastasis and lung metastasis (Fig. 2M,N), as well as patients who did not undergo chemotherapy (Fig. 2I) or radiotherapy (Fig. 2N,J).

Correlations among variables

Before we performed the Cox regression analysis, Spearman's correlation was used to ensure that there was no collinearity existed between screened variables. The results of correlation analyses are presented in Fig. 3.

Nomogram variable screening

According to the Cox regression results, the model containing age, grade, surgery, chemotherapy, lung metastasis and bone metastasis had minimal P value in the training cohort. In the univariate regression analysis, nine variables (age, grade, T stage, N stage, chemotherapy, surgery, bone metastasis, lung metastasis and primary site) were significantly associated with OS. Multivariate Cox regression analysis showed that age (>65 years old) ($P = 0.003$, hazard ratios (HR) = 1.194, 95% confidence interval (CI) = 1.061–1.343), grade III ($P < 0.001$, HR = 1.456, 95% CI = 1.287–1.648) and chemotherapy ($P < 0.001$, HR = 0.268, 95% CI = 0.235–0.305) were independent prognostic factors in patients with GCLM. More details are presented in Table 3.

For the grouping status of CSS, the detailed information of patients with GCLM in the CSS group is shown in Table 4. Univariate Cox regression analysis demonstrated that sex, race, grade, N stage, chemotherapy, surgery, lung metastasis and bone metastasis were CSS-related prognostic factors. Then the multivariate Cox regression analysis was carried out to screen the independent prognostic factors in patients

with GCLM. These results were tabulated in Table 4.

Nomogram construction and validation

We constructed two nomograms for GCLM patients according to the variables screen. Fig. 4A shows an example of using the nomogram to predict the overall survival probability of a given patient. And Fig. 4B shows a competing event nomogram to assess the 1- and 3-year chances of CSS by incorporating those independent prognostic predictors. The likelihood that other causes contributed to or directly caused the death of gastric cancer was assessed via this model by computing the total score by each of the measured individual variables. And the calibration curves of these nomograms showed optimal consistency of the actual likelihood with the nomogram-forecasted likelihoods in both the training and validation cohort (Fig. 5).

Fig. 6A,B depict the AUC in time-dependent AUC curves in the Cox model and the Competing risk model, respectively. It was greater than 0.7 for the prediction of OS and CSS within three years, indicating that the nomogram had good discrimination. In the training cohort, the AUCs of the Cox model for forecasting 1 and 3 years were 0.794 and 0.761, respectively (Fig. 6C). The AUCs at 1 and 3 years in the validation cohort were 0.739 and 0.794, respectively (Fig. 6D). The results showed that the AUCs of the competing risk model was 0.726 at one year and 0.734 at three years, while in the verification group, the AUC was 0.779 at one year and 0.826 at three years (Fig. 6E,F).

The decision curve analysis was performed on the training and verification cohort, and the results are shown in Fig. 7. Besides that, the plots for 1-year and 3-year survival showed good net benefits, indicating the nomogram's superior prediction accuracy.

Clinical value comparison between nomograms and AJCC stage system

To further estimate the clinical usefulness of nomograms in this study, we used NRI, IDI and C-Index to compare the accuracy between the nomogram and the AJCC stage system. While using the nomogram in the training cohort, the C-index was 0.7403 (95% CI = 0.7259–0.7546) in the nomogram and 0.5724 (95% CI = 0.5536–0.5912) in the AJCC Stage system, the NRI for the 1- and 3-year OS were 0.7870 (95% CI = 0.6564–0.9210) and 0.6991 (95% CI = 0.4523–0.8901), and the IDI values for 1- and 3-year OS were 0.184 (95% CI = 0.156–0.212, $P < 0.001$) and 0.095 (95% CI = 0.067–0.132, $P < 0.001$) (Table 5). These results were verified in the validation cohort (Table 5), indicating that the nomogram predicted prognosis with greater accuracy than the AJCC Stage System.

Risk stratification for gastric cancer patients with liver metastasis

We calculated total scores based on the nomogram for risk stratification. The patient was therefore classified into two risk groups (low-risk and high-risk) based on the median risk score. The Kaplan-Meier survival analysis revealed that the low-risk group had a better OS and CSS than the high-risk group (Fig. 8C–F). Meanwhile, the nomogram demonstrated excellent discrimination between two risk groups, whereas the AJCC Stage System had limited ability to distinguish between these two groups (Fig. 8A,B). Both results were confirmed in the validation cohort.

Discussion

Gastric cancer is one of the most common gastrointestinal tract malignant tumors with a low early diagnosis rate, low surgical resection rate, and high recurrence [18]. A significant proportion of gastric cancer patients had distant metastasis and the most frequent sites of gastric cancer are liver, peritoneal and distant lymph nodes [19]. These patients usually have a poor prognosis partially owing to the lack of effective treatment. In our study, we constructed two nomograms to predict the

Table 3
Univariate and multivariate Cox proportional hazards regression analysis of patients with GCLM in the OS group.

Variables	Univariate Cox regression analysis		Multivariate Cox regression analysis			
HR	95% CI	P	HR	95% CI	P	
age						
<65	1 (reference)			1 (reference)		
>65	1.359	1.215–1.521	0.000	1.194	1.061–1.343	0.003
Sex						
female	1 (reference)					
male	1.124	0.994–1.272	0.062			
Race						
American Indian/Alaska Native	1 (reference)					
Asian/Pacific Islander	0.572	0.301–1.087	0.088			
Black	0.592	0.314–1.118	0.106			
White	0.605	0.324–1.129	0.114			
Grade						
I	0.682	0.483–0.965	0.030	0.566	0.399–0.805	0.002
II	1 (reference)			1 (reference)		
III	1.382	1.225–1.561	0.000	1.456	1.287–1.648	0.000
IV	0.683	0.483–0.965	0.331	1.416	0.906–2.211	0.127
T Stage						
T0	1 (reference)			1 (reference)		
T1	4.165	0.585–29.65	0.154	4.816	0.671–34.577	0.118
T2	2.921	0.405–21.06	0.288	4.142	0.569–30.161	0.161
T3	3.052	0.428–21.76	0.266	4.502	0.625–32.459	0.135
T4	3.960	0.556–28.21	0.170	5.146	0.714–37.076	0.104
N Stage						
N0	1 (reference)			1 (reference)		
N1	0.865	0.767–0.977	0.019	0.910	0.802–1.033	0.146
N2	0.714	0.576–0.886	0.002	0.845	0.672–1.063	0.151
N3	0.836	0.674–1.036	0.102	1.188	0.940–1.501	0.150
Radiation						
Yes	0.919	0.796–1.061	0.252			
No	1 (reference)					
Chemotherapy						
Yes	0.329	0.293–0.370	0.000	0.268	0.235–0.305	0.000
No	1 (reference)			1 (reference)		
Surgery						
Yes	0.583	0.496–0.685	0.000	0.476	0.394–0.574	0.000
No	1 (reference)			1 (reference)		
Bone Metastasis						
Yes	1.250	1.026–1.524	0.027	1.256	1.025–1.540	0.028
No	1 (reference)			1 (reference)		
Brain Metastasis						
Yes	1.368	0.808–2.318	0.244			
No	1 (reference)					
Lung Metastasis						
Yes	1.526	1.312–1.775	0.000	1.511	1.293–1.766	0.000
No	1 (reference)			1 (reference)		
Marital						
Yes	0.965	0.860–1.082	0.540			
No	1 (reference)					
Primary site						
Cardia	1.050	0.889–1.240	0.564	1.096	0.920–1.306	0.304
Pylorus	0.982	0.598–1.614	0.943	0.886	0.537–1.462	0.637
Body	1.141	0.904–1.441	0.266	1.044	0.825–1.322	0.719
Antrum	1 (reference)			1 (reference)		
Fundus	1.240	0.930–1.653	0.142	1.079	0.806–1.445	0.610
Lesser curve	1.043	0.798–1.363	0.00475	0.958	0.732–1.254	0.754
Greater curve	1.414	1.015–1.969	0.040	1.211	0.864–1.696	0.267
other	1.306	1.075–1.587	0.007	1.108	0.910–1.350	0.308
Histologic type						
Adenocarcinoma	1 (reference)					
Signet ring cell	1.152	0.928–1.429	0.200			
Other	1.057	0.876–1.275	0.566			

prognosis of GCLM patients. And the validation of these nomograms showed that they had good discriminative performance and calibration. In addition, the risk stratification also showed a favorable ability to categorize GCLM patients into high- and low-risk groups with significant differences.

Previous research suggests that some factors, such as differentiation, clinical phenotype, and adjuvant therapies, may potentially affect the survival of patients with GCLM. As a result, we included as many of these factors as possible in the Cox and competing risk model. According to Hu et al, differentiation degree was related to OS, and poorly differentiated

adenocarcinoma was confirmed as a high-risk factor for GCLM [20]. Similar results were observed in our study, with GCLM patients with poor differentiation (Grade III) receiving the highest score. Different standards exist in the clinical phenotype of gastric cancer, and the WHO classification is now used globally. Although common types, such as papillary carcinoma and Signet ring cell carcinoma, are easily agreed upon, the specialized types remain contentious [21]. Cox regression analysis revealed no correlation between the prognosis and pathological types of GCLM patients, according to Li et al. [22]. In contrast, Daniela et al found pretty similar survival values for patients with various types

Table 4
Results of univariate and multivariate analyses by Fine-Gray proportional subdistribution hazards model.

Variables	Univariate Cox regression analysis			Multivariate Cox regression		
	HR	95% CI	P	HR	95% CI	P
age						
<65	1 (reference)					
>65	1.048	0.927–1.184	0.454			
Sex						
female	1 (reference)			1 (reference)		
male	1.157	1.005–1.332	0.043	1.147	0.980–1.343	0.088
Race						
American Indian/Alaska Native	1 (reference)			1 (reference)		
Asian/Pacific Islander	0.610	0.246–1.513	0.286	0.892	0.370–2.147	0.798
Black	0.391	0.159–0.959	0.004	0.626	0.264–1.486	0.288
White	0.502	0.207–1.222	0.129	0.702	0.300–1.642	0.414
Grade						
I	0.540	0.390–0.748	0.000	0.453	0.327–0.630	0.000
II	1 (reference)			1 (reference)		
III	1.214	1.064–1.385	0.0000	1.181	1.026–1.359	0.020
IV	0.893	0.512–1.557	0.0441	0.940	0.532–1.660	0.830
T Stage						
T0	1 (reference)					
T1	2.864	0.197–41.73	0.441			
T2	2.902	0.197–42.66	0.437			
T3	2.884	0.198–42.04	0.438			
T4	2.991	0.205–43.64	0.423			
N Stage						
N0	1 (reference)			1 (reference)		
N1	1.215	1.062–1.391	0.005	1.240	1.070–1.438	0.004
N2	1.006	0.795–1.273	0.961	1.214	0.946–1.558	0.128
N3	1.104	0.866–1.403	0.423	1.526	1.171–1.989	0.002
Radiation						
Yes	0.974	0.844–1.124	0.716			
No	1 (reference)					
Chemotherapy						
Yes	0.578	0.490–0.681	0.000	0.459	0.384–0.548	0.000
No	1 (reference)			1 (reference)		
Surgery						
Yes	0.581	0.486–0.693	0.000	0.516	0.420–0.635	0.000
No	1 (reference)			1 (reference)		
Bone Metastasis						
Yes	1.315	1.078–1.603	0.007	1.062	0.848–1.330	0.599
No	1 (reference)			1 (reference)		
Brain Metastasis						
Yes	1.163	0.673–2.010	0.588			
No	1 (reference)					
Lung Metastasis						
Yes	1.717	1.444–2.041	0.000	1.583	1.315–1.906	0.000
No	1 (reference)			1 (reference)		
Marital						
Yes	1.008	0.882–1.151	0.908			
No	1 (reference)					
Primary site						
Cardia	0.991	0.726–1.277	0.927			
Pylorus	0.832	0.489–1.417	0.498			
Body	0.963	0.726–1.277	0.795			
Antrum	1 (reference)					
Fundus	1.117	0.813–1.534	0.496			
Lesser curve	0.864	0.638–1.169	0.342			
Greater curve	1.138	0.720–1.799	0.580			
other	1.052	0.823–1.343	0.687			
Histologic type						
Adenocarcinoma	1 (reference)					
Signet ring cell	1.351	1.073–1.702	0.011			
Special type	1.047	0.824–1.331	0.708			

of adenocarcinomas, but carcinomas with "signet-ring" cells proved to be extremely aggressive, which was consistent with our findings [23]. In randomized controlled trials, the clinical benefit of chemotherapy in patients with advanced gastric cancer has been demonstrated in randomized controlled trials [24–26]. Chemotherapy is currently the mainstay of therapy for GCLM patients, expected to alleviate disease-related symptoms and prolong survival [27].

Another unexpected factor that deserves special mention is surgery. The surgical treatment of patients with distant metastases has remained contentious in recent years. Systemic chemotherapy is currently the

standard management for GCLM, but it does not produce satisfactory results [28]. For patients with incurable factors such as unresectable liver metastasis, the Japanese Guidelines (5th edition) strongly recommend not to perform gastrectomy as a reduction surgery. However, surgery such as hepatectomy is recommended in patients with limited metastasis where there is no other incurable factor [27]. A retrospective study reported by Sheraz R Marker et al, revealed hepatectomy for synchronous gastric cancer liver metastases may improve survival benefits in certain patients [29]. On the contrary, the European Society for Medical Oncology (ESMO) reported that metastasis resection

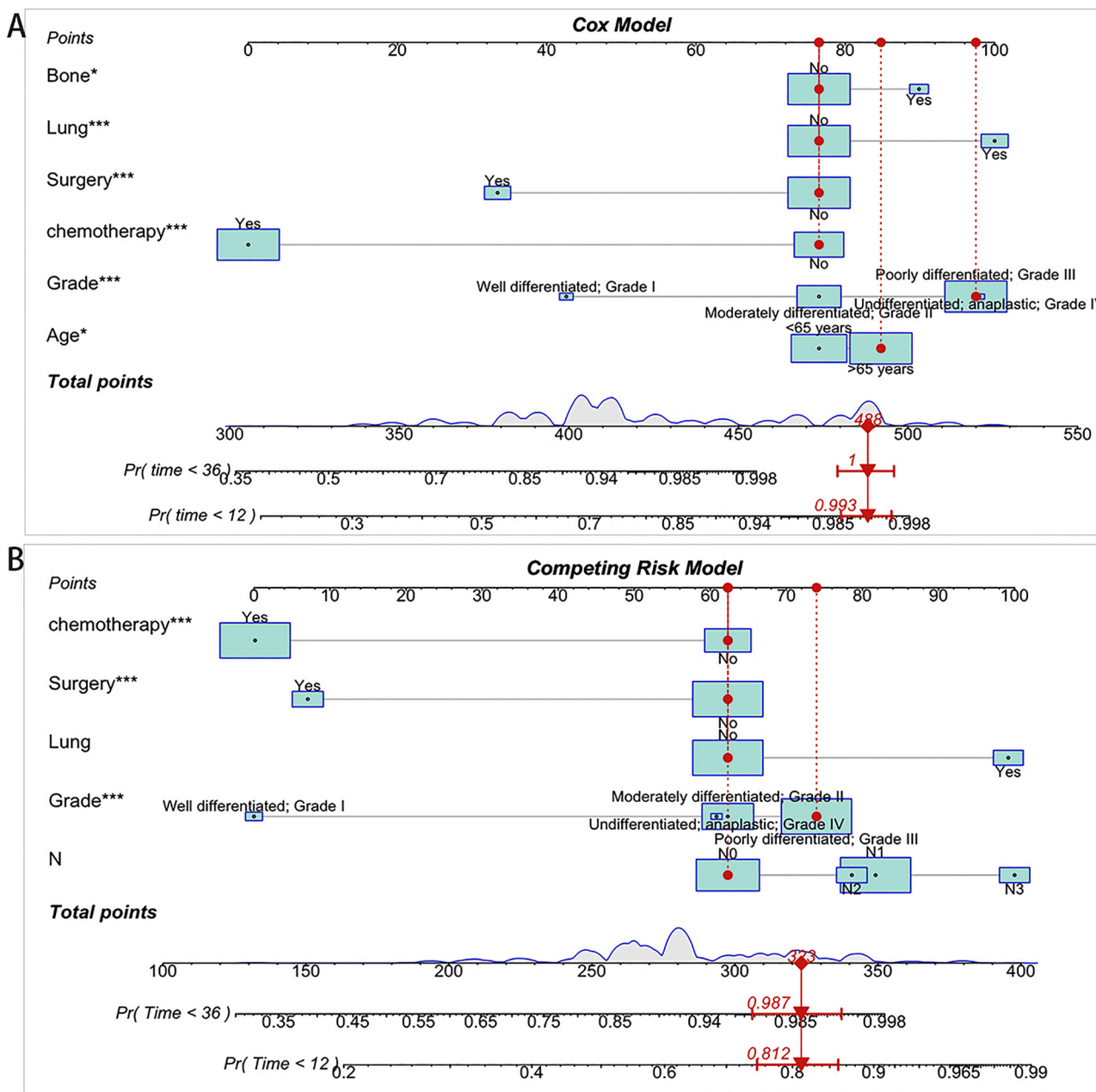


Fig. 4. Constructed nomograms for prognostic prediction of overall survival and cancer-specific survival. (A) Nomogram for overall survival in GCLM patients. (B) Nomogram for cancer-specific survival in GCLM patients.

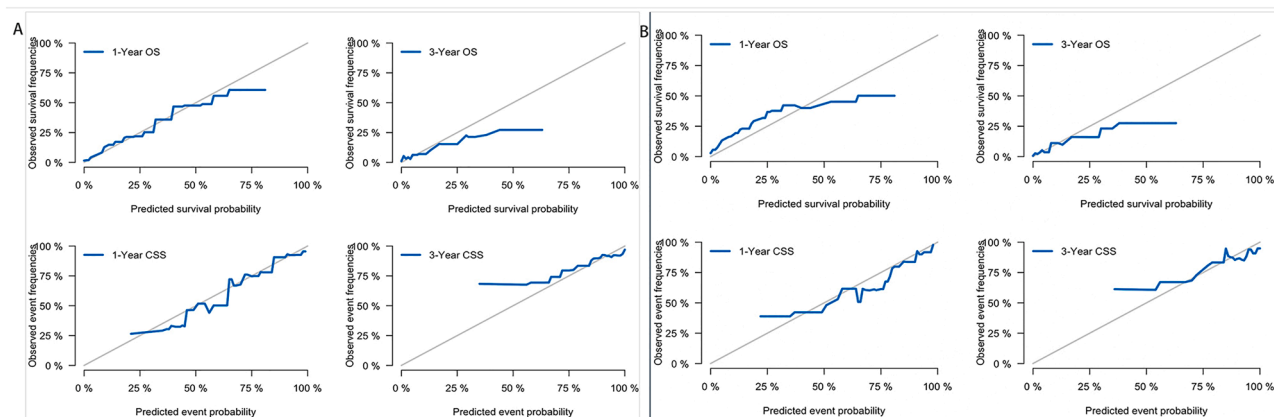


Fig. 5. Calibration curves. (A) 1-year and 3-year likelihoods of OS and CSS in the training dataset. (B) 1-year and 3-year likelihoods of OS and CSS in the validation dataset.

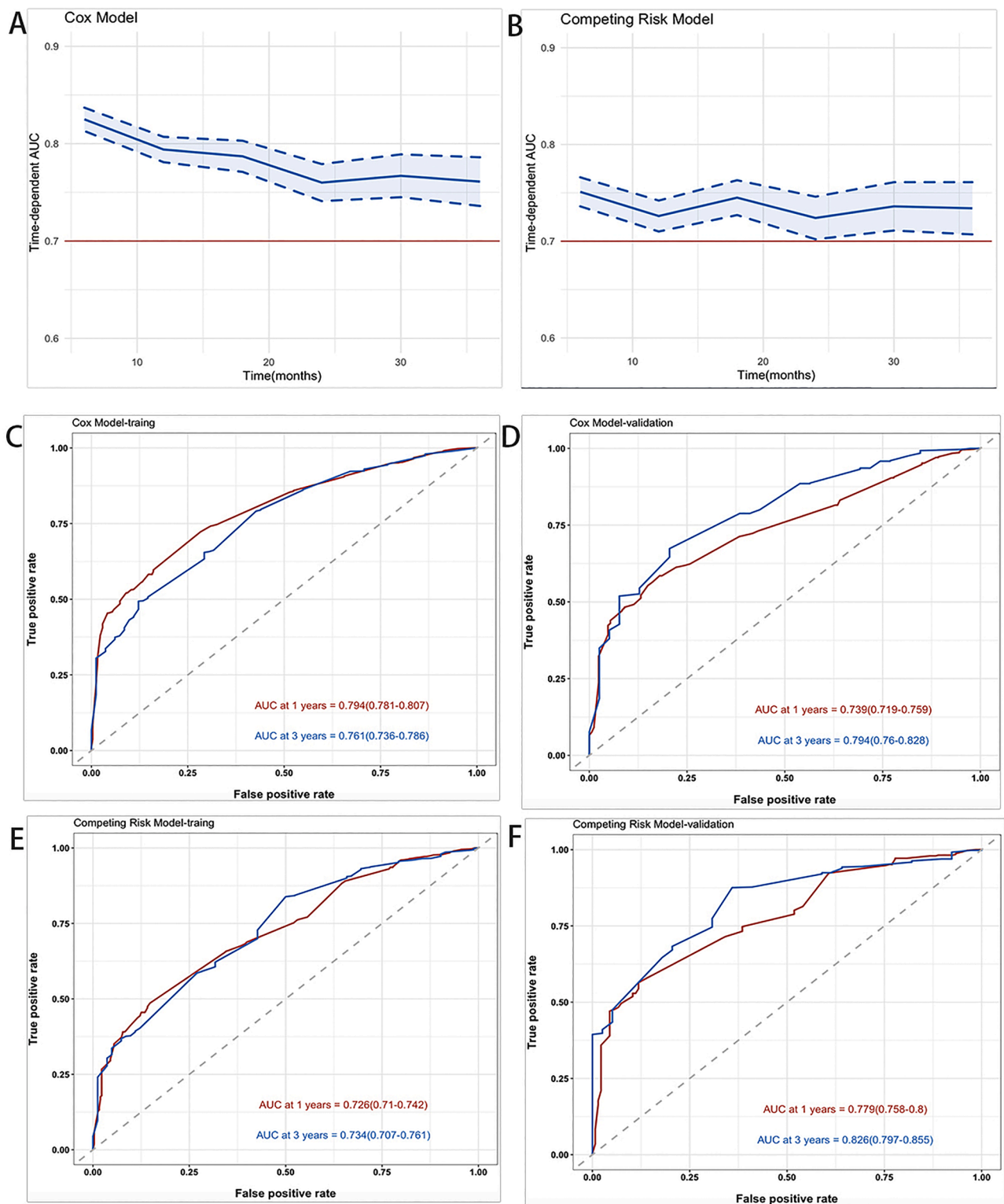


Fig. 6. Time-dependent AUC and receiver operating characteristic (ROC) curves of OS and CSS. (A,B) Time-dependent AUC of using the nomogram to OS and CSS probability within 3 years in the training cohort and validation cohort. The blue line represents AUC = 0.7, which is considered ideal. And the shading area between blue dotted curves represents 95% credible intervals. (C,D) ROC curves corresponding to 1-year and 3-year OS in the training and validation cohort, respectively. (E, F) ROC curves corresponding to 1-year and 3-year CSS in the training and validation cohort, respectively.

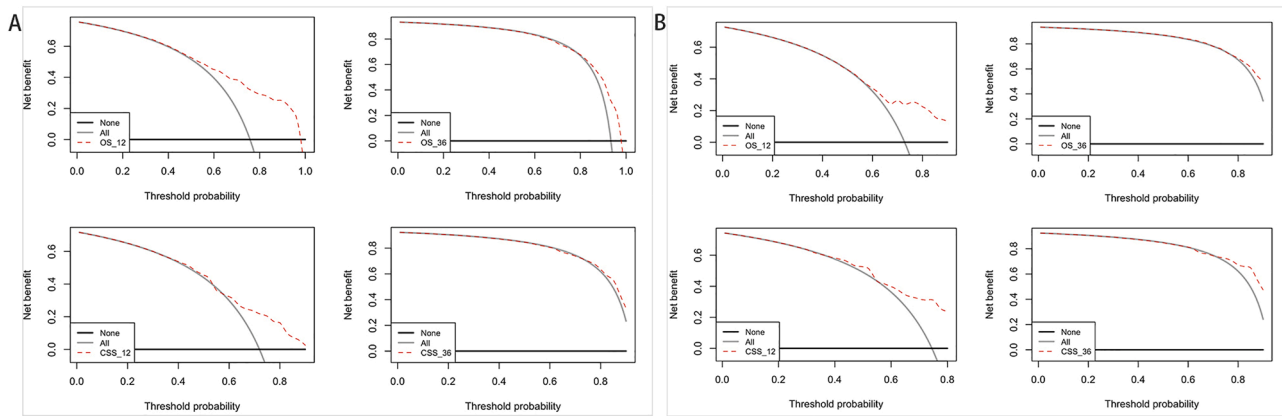


Fig. 7. Decision curve analysis of the nomogram in the estimation of OS and CSS of patients with GCLM. (A) Training cohort. (B) Validation cohort.

Table 5
Comparison of different models for estimating the overall survival of GCLM patients.

Training cohort				Validation cohort			
Index	Estimate	95%CI	P value	Estimate	95%CI	P value	
NRI (vs. AJCC stage System)							
For 1-year OS	0.7870	0.6564–0.9210		0.6826	0.4294–0.8421		
For 3-year OS	0.6991	0.4523–0.8901		0.5485	0.2511–0.9997		
IDI (vs. AJCC stage System)							
For 1-year OS	0.184	0.156–0.212	<0.0001	0.156	0.116–0.195	<0.0001	
For 3-year OS	0.095	0.067–0.132	<0.0001	0.108	0.061–0.169	<0.0001	
C-Index							
The nomogram (OS)	0.7403	0.7259–0.7546		0.6953	0.6728–0.7179		
AJCC Stage System	0.5724	0.5536–0.5912		0.5495	0.5220–0.5769		

does not benefit patients with metastatic disease [30]. The clinical trial (REGATTA) identified that gastrectomy does not improve survival in patients with limited metastasis [31]. However, the guidelines also stated that after palliative chemotherapy, the possibility of surgery should be reconsidered. According to Al-Batran SE et al, gastric cancer patients with limited metastases who received neoadjuvant chemotherapy and underwent surgery had a favorable survival [32]. The purpose of this study was to determine whether surgery was a significant factor in survival in GCLM. Based on the univariate and multivariate Cox regression results ($P < 0.001$), we concluded that surgery affects the survival of patients with GCLM and was mentioned as a possible factor in the nomogram’s establishment.

We identified that the surgery improved patients’ survival based on the prediction model. Even though, various factors such as histology, depth of invasion, lymphatic or venous invasion, number and size of liver metastases, surgical options such as LR only, TG+LR, or TG only, and so on may be associated with the outcomes of undergoing surgery for patients with GCLM [27]. To the best of our knowledge, no other studies have used the SEER database to determine that surgery is an independent prognostic factor that has a positive impact on the survival of GCLM patients. Our research does, however, have some limitations. This study was not a randomized controlled trial, and selection bias in the cohort was unavoidable. Additionally, due to the small number of patient’s limitations and incomplete information, some important indicators, such as the size and number of liver metastases, cannot be evaluated. As a result, more in-depth and comprehensive studies may be required.

The analysis of GCLM in the current study provided an opportunity to reconsider some factors that could be incorporated into the prognostic nomogram. Our nomograms incorporate more factors, such as demographic characteristics than the traditional AJCC staging system, and they are more accurate in predicting patient outcomes and assisting clinical practice. The AJCC Stage System has traditionally been the first choice for predicting the prognosis of gastric cancer patients. There are

several nomograms for GC that have been shown to have clinical utility. Memorial Sloan-Kettering Cancer Center’s Dikken JL et al established a nomogram to predict the survival of gastric cancer patients after an R0 resection [33]. Furthermore, the collagen nomogram and radionics nomogram make significant advances in the diagnosis and evaluation of recurrence and metastasis [34,35]. With today’s technology, precision medicine is becoming more feasible. With the rapid progress of genomics, metabolomics, and radiomics, multi-omics analysis is becoming ever more popular. In the future, the physician will collect as much information about patients as possible during their visits, resulting in an exhaustive analysis of the various dimensions. We believe that these studies will provide a firm foundation for personalized treatment to improve GC patient life expectancy.

Despite the fact that the nomogram performed well, the current study had some weaknesses. The SEER database information, such as surgical methods and specific chemotherapy regimens, is insufficient. These factors may have an impact on the accuracy of those patients’ prognoses. Further to that, it is unknown whether the patients have synchronous or metachronous liver metastases, and the medical management for the two groups of patients differs. What is more, a large proportion of gastric cancer patients have a signet-ring cancer phenotype, which is more likely to metastasize to the ovaries and peritoneum [36]. The database, however, only contains information on four types of metastases: liver, lung, bone, and brain. Our findings were derived from a cohort of Americans. As a result, a larger-sample multicenter study should be conducted to determine whether our study results are more broadly applicable.

Conclusion

We created two prognostic nomograms and a risk stratification system using the SEER database. It also laid the foundation for precision therapy and tailor-made treatment in GCLM patients. The optimal surgical procedure and conditions for GCLM should be studied further.

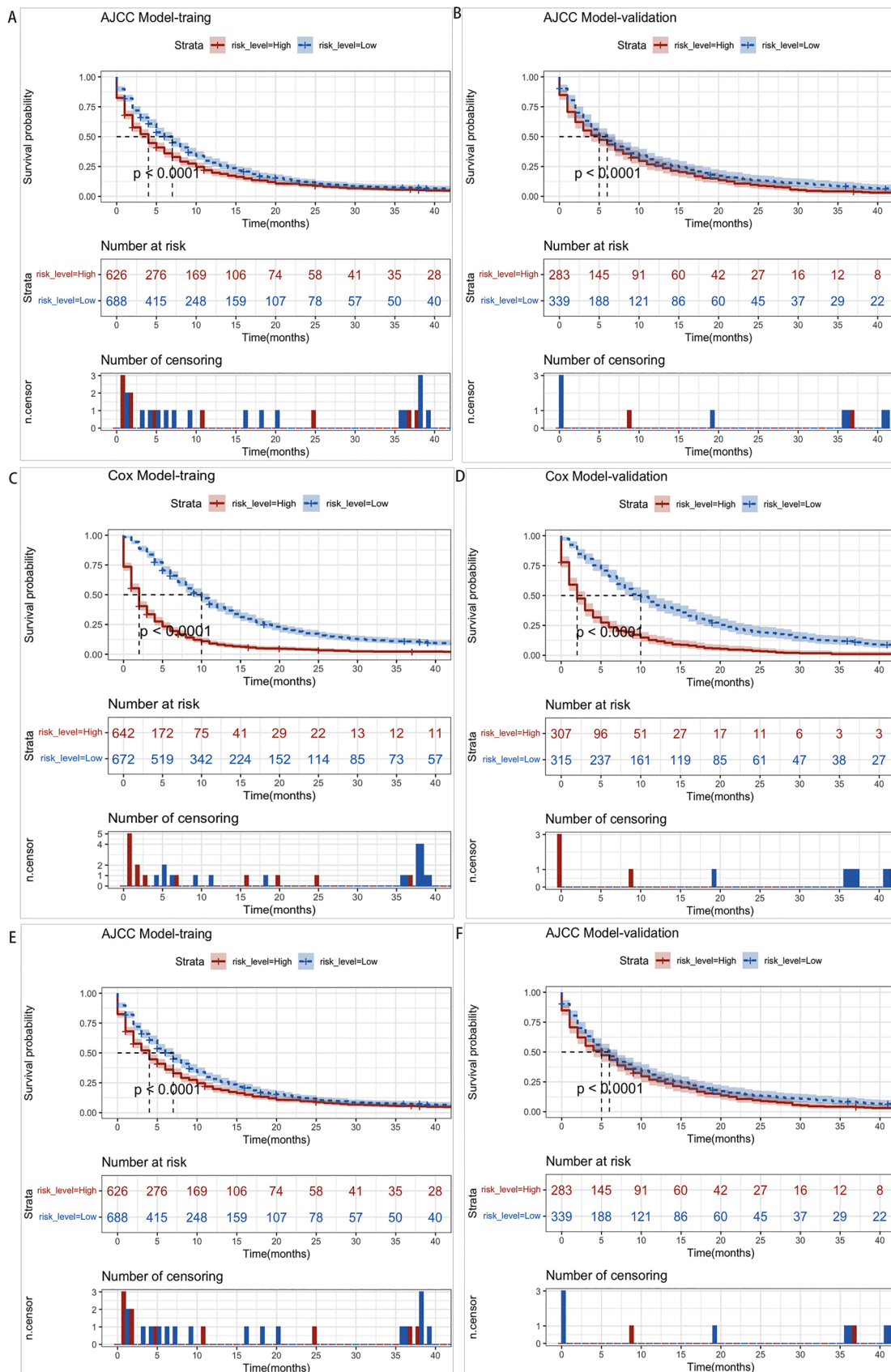


Fig. 8. Kaplan–Meier OS and CSS curves of GCLM patients with different risks stratified by the nomogram. (A,B) GCLM patients in the training and validation cohort at different stages are classified according to the AJCC staging system. (C,D) GCLM patients in the training and validation cohort at different stages are classified according to the cox model nomogram. (E,F) GCLM patients in the training and validation cohort at different stages are classified according to the competing risk model nomogram.

CRedit authorship contribution statement

Zhongyi Dong: Conceptualization, Methodology, Software, Writing – original draft. **Yeqian Zhang:** Data curation. **Haigang Geng:** Software, Writing – original draft. **Bo Ni:** Visualization. **Xiang Xia:** Investigation. **Chunchao Zhu:** Supervision. **Jiahua Liu:** Validation, Writing – review & editing. **Zizhen Zhang:** Writing – review & editing.

Declaration of Competing Interest

All authors declare that they have no competing interests.

Acknowledgments

We kindly thank all the staff of National Cancer Institute for their efforts toward the SEER program.

Funding

The study was sponsored by The National Natural Science Foundation of China (Grant Nos. 81972206 and 82173215)

Availability of data and materials

All data generated during this study are included in this published article. Further inquiry data can be obtained by the corresponding author.

Ethics approval and consent to participate

Not applicable.

Consent for publication

Not applicable.

Publisher's note

Springer Nature remains neutral with regard to jurisdictional claims in published maps and institutional affiliations.

Supplementary materials

Supplementary material associated with this article can be found, in the online version, at doi:10.1016/j.tranon.2022.101480.

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