ORIGINAL RESEARCH

Development of Prognostic Nomogram Based on Lipid Metabolic Markers and Lactate Dehydrogenase in Non-Metastatic Nasopharyngeal Carcinoma

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Purpose: To establish and verify a comprehensive prognostic nomogram for predicting survival outcomes and improving the prognosis for non-metastatic nasopharyngeal carcinoma (NPC).

Patients and Methods: Our retrospective study screened 613 cases of non-metastatic NPC who received radiotherapy from July 2012 to December 2016. A reliable nomogram was formulated for predicting overall survival (OS) and progression-free survival (PFS) using all independent predictors selected by Cox regression analysis. A comparison is conducted between the current staging and the predictive performance of the nomogram. Internal validation was performed in a single center using the validation cohort to assess predictive accuracy and discrimination.

Results: High-density lipoprotein cholesterol, Epstein–Barr virus DNA and lactate dehydrogenase were determined to be valuable predictive indicators for predicting OS and PFS. Triglycerides were a valuable predictive indicator for predicting OS. Calibration curves demonstrated that the nomogram had remarkable correspondence between the prediction outcomes and the actual observations. Receiver operating characteristic curves showed that the nomogram had greater area under the curve and more satisfactory discrimination capability than the current TNM staging. Decision curve analysis revealed that the nomogram had high net clinical benefits. Significant differences were observed when low- and high-risk groups were stratified via Kaplan–Meier curves.

Conclusion: Our proposed nomogram combining lipid metabolic markers and lactate dehydrogenase could assist clinicians in the accurate prognostic prediction of non-metastatic NPC patients and provide personalized treatment recommendations.

Keywords: nomogram, lipid metabolic markers, nasopharyngeal carcinoma, lactate dehydrogenase, triglycerides

Introduction

Nasopharyngeal carcinoma (NPC) is an insidious cancer of the head and neck, with the most frequent occurrence in East and South Asia, commonly arising in the pharyngeal recess.¹ Worldwide, about 133,000 newly diagnosed cases of NPC and 80,000 deaths occur annually from the latest statistical data in 2020.² Radiotherapy (RT) with or without chemoradiotherapy is the mainstay of treatment for NPC due to its radiosensitive and chemosensitive nature.³ In recent years, the rate of locoregional recurrence control has increased considerably since the incorporation of intensity-modulated radiotherapy (IMRT) and chemotherapy.⁴ However, despite receiving IMRT, the 5-year overall survival (OS) remains 80%–87%, with recurrence and distant failure occurring at an incidence of 5%–15% and 15%–30%, respectively.^{5–7} Hence, the treatment of NPC remains a massive challenge until now. The development of an effective tool for determining whether patients are at high risk of mortality and progression is urgently needed.

At present, the prognosis prediction and clinical treatment strategy formulation are determined by the tumor-node-metastasis (TNM) staging. Because the TNM staging solely describes the anatomical range of tumors, there may be diversity in survival

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Recently, the concept of nasopharyngeal cancer ecology has been proposed, considering cancer as an ecological evolutionary process. The growth of caner cells depends on substances in the tumor microenvironment (TME), both of which exchange substances to achieve mutualism and co-evolve, while the progression of malignancy is regarded as ecological invasion.¹⁰ To adapt to the new TME, the metabolism of cancer cells is reprogrammed to support malignant proliferation.¹¹ In particular, lipid metabolism contributes to the thriving of cancer cells, providing advantages for metastasis.¹² Accumulating evidence confirms that lipid metabolism and inflammation are closely relevant to tumor survival, migration, invasion and metastasis.^{13–16} To date, lipid metabolism markers, such as total cholesterol (TC), triglycerides (TG), high-density lipoprotein cholesterol (HDL-C), lowdensity lipoprotein cholesterol (LDL-C), apolipoprotein A1 (Apo A1) and apolipoprotein B (Apo B), have been found to be associated with prevalence and prognosis in many malignancies, including breast cancer, colorectal neoplasms, esophageal squamous cell carcinoma, and non-small cell lung cancer.¹⁷⁻²⁰ Some retrospective studies correlate levels of TG to the progression of NPC.^{21,22} HDL-C has been illustrated as a potential prognostic indicator for NPC, with low levels associated with a worse outcome.²³ Additionally, many inflammatory biomarkers are proven to supply effective prediction guidance for NPC patients, such as lactate dehydrogenase (LDH), platelet-to-lymphocyte ratio (PLR), neutrophil-to-lymphocyte ratio (NLR) and lymphocyte-monocyte ratio (LMR).²⁴⁻²⁶ However, the connection between lipid metabolism markers and prognosis in NPC patients has been investigated in few studies. Only one or two markers were examined without combining inflammatory biomarker. Thus, in this study, we integrated lipid metabolism markers and inflammatory index to establish a nomogram with more accurate prediction of survival events than current TNM staging systems for non-metastatic NPC.

Methods

Patients

Patients treated at Guangxi Medical University Cancer Hospital from July 2012 to December 2016 were reviewed in our retrospective observational research. In order to explore early-, middle- and advanced-stage patients more comprehensively, we selected patients with non-metastatic NPC as the study population. The eligibility criteria included (1) histologically confirmed stage I–IVa NPC by the 8th edition TNM staging system; (2) adequate clinical information and follow-up data; (3) no prior other malignancies; (4) receiving IMRT. Exclusion criteria included (1) taking lipid-modulating medicine regularly before treatment; (2) receiving anticancer treatment in the past; (3) lack of serum lipid data. Overall, we randomly assigned 613 patients who met the criteria to the training cohort (429 patients) and validation cohort (184 patients) in a 7:3 ratio. The study was approved by the Ethics Committee of Guangxi Medical University Cancer Hospital in China (IRB approval number: LW2023023) and complied with the Declaration of Helsinki. Informed consent was waived for this study because of its retrospective nature. All patient data were anonymized.

Baseline Data Collection

Before initiation of treatment, we collected the baseline clinical information from the electronic medical records system, including age, gender, height, weight, smoking status, TNM stage, and therapeutic regimen. The following laboratory data were acquired within 2 weeks pre-treatment: plasma EBV DNA load, albumin (ALB), hemoglobin (HGB), inflammatory biomarkers (LDH, PLR, NLR and LMR), lipid metabolism markers (TC, TG, HDL-C, LDL-C, Apo A1 and Apo B). PLR was calculated as dividing the absolute platelet count by the absolute lymphocyte count. Dividing the absolute neutrophil count by the absolute lymphocyte count divided by the absolute monocyte count is LMR.

Treatment

All eligible patients finished IMRT with a total dose of 70–74.6Gy/30-33 fractions to the primary gross tumor volume as established radiotherapy schedule, while the cervical metastatic lymph node was given 60–72.6Gy/30-33 fractions, 5 times per week. The high-risk clinical target volume and the low-risk clinical target volume were irradiated with 60–62.7Gy/30-33 fractions and 52–56.1Gy/30-33 fractions, respectively. Treatment regimens of NPC patients were RT alone for stage I, RT alone or concurrent chemoradiotherapy (CCRT) for stage II, and CCRT or induction chemotherapy (IC) + CCRT or CCRT + adjuvant chemotherapy (AC) for stage III–IVA. Patients received 1–3 cycles of cisplatin as CCRT regimen every 21 days. Patients received IC and AC regimens including GP regimen (1000 mg/m² of gemcitabine on days 1 and 8, 80 mg/m² of cisplatin on day 1), TPF (60 mg/m² of docetaxel on day 1, 60 mg/m² cisplatin on day 1, 3000 mg/m² of 5-fluorouracil on days 1–5), TP (75 mg/m² of docetaxel on day 1, 75 mg/m² of cisplatin on day 1), or PF (80 mg/m² of cisplatin on day 1, 750 mg/m² of 5-fluorouracil on days 1–5) for 1–3 cycles every 21 days. Patients who could not tolerate cisplatin were given carboplatin, nedaplatin or loplatin.

Follow-Up and End Point

Patients returned to the outpatient department for reexamination every 3 months for the first 2 years, every 6 months for the next 3 years, and once a year after 5 years after treatment. The contents of the reexamination included physical examination, hematology examination, nasopharyngeal fiberscope, computerized tomography of the chest and abdomen, magnetic resonance imaging of the head and neck, and bone emission computed tomography if necessary. Telephone follow-up was used for patients who did not return as planned. The primary endpoint in our study was OS, which was calculated from the time from diagnosis on to death or last follow-up. The secondary endpoint was progression-free survival (PFS), which was calculated from the time from diagnosis to death, the occurrence of treatment failure or last follow-up.

Statistical Analysis

Differences between continuous variables were assessed using Mann–Whitney *U*-tests, while differences between categorical variables were evaluated using chi-square tests. In the training cohort, continuous variables applied the function offered in the R package "survminer" to obtain the optimal cut-off values, which were converted to categorical variables based on the calculated cut-off values. Independent prognostic factors were acquired by multivariate Cox regression analysis (P < 0.05) based on those factors that showed P < 0.05 in univariate analysis. The nomogram was formulated to predict survival events of patients based on all included independent prognostic factors using R package "rms". The actual OS and PFS were compared with the predicted OS

Variables	Overall Training Cohort		Validation Cohort	P-value
	(n = 613)	(n = 429)	(n = 184)	
Age (mean (SD))	47.74 (10.55)	47.80 (10.53)	47.58 (10.61)	0.8108
Gender (%)				0.8179
Female	171 (27.90)	118 (27.51)	53 (28.80)	
Male	442 (72.10)	311 (72.49)	131 (71.20)	
Smoking (%)				0.5477
No	404 (65.91)	279 (65.03)	125 (67.93)	
Yes	209 (34.09)	150 (34.97)	59 (32.07)	
T stage (%)				0.3276
TI/T2	250 (40.78)	169 (39.39)	81 (44.02)	
T3/T4	363 (59.22)	260 (60.61)	103 (55.98)	

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Variables	Overall	Training Cohort	Validation Cohort	P-value
	(n = 613)	(n = 429)	(n = 184)	
N stage (%)				0.9776
N0/N1	352 (57.42)	247 (57.58)	105 (57.07)	
N2/N3	261 (42.58)	182 (42.42)	79 (42.93)	
TNM stage (%)				0.1090
I	13 (2.12)	6 (1.40)	7 (3.80)	
II	140 (22.84)	92 (21.45)	48 (26.09)	
III	258 (42.09)	189 (44.06)	69 (37.50)	
IVa	202 (32.95)	142 (33.10)	60 (32.61)	
EBV DNA (%)				0.7484
≤7960	460 (75.04)	324 (75.52)	136 (73.91	
>7960	153 (24.96)	105 (24.48)	48 (26.09)	
Treatment (%)				0.2298
RT	66 (10.77)	39 (9.09)	27 (14.67)	
CCRT	300 (48.94)	215 (50.12)	85 (46.20)	
IC+CCRT	203 (33.12)	143 (33.33)	60 (32.61)	
CCRT+AC	44 (7.18)	32 (7.46)	12 (6.52)	
ALB (mean (SD))	42.95 (3.81)	42.90 (3.97)	43.05 (3.42)	0.6578
BMI (median [IQR])	22.32 [20.68, 24.46]	22.27 [20.57, 24.54]	22.49 [20.74, 24.40]	0.8635
HGB (median [IQR])	140.00 [126.00, 149.00]	139.00 [126.00, 149.00]	140.00 [129.00, 148.00]	0.9593
LDH (median [IQR])	176.00 [153.00, 203.00]	177.00 [153.00, 202.00]	171.50 [153.75, 207.25]	0.6977
TC (median [IQR])	5.01 [4.38, 5.79]	5.00 [4.38, 5.74]	5.10 [4.39, 5.89]	0.3145
TG (median [IQR])	1.25 [0.93, 1.90]	1.28 [0.93, 1.97]	1.20 [0.93, 1.74]	0.1912
HDL-C (median [IQR])	1.26 [1.12, 1.44]	1.25 [1.10, 1.44]	1.27 [1.13, 1.45]	0.5415
LDL-C (median [IQR])	3.27 [2.74, 3.93]	3.24 [2.75, 3.91]	3.345 [2.72, 3.99]	0.2748
Apo AI (median [IQR])	1.24 [1.09, 1.42]	1.24 [1.10, 1.41]	1.25 [1.08, 1.45]	0.8862
Apo B (median [IQR])	0.97 [0.82, 1.14]	0.97 [0.82, 1.13]	0.98 [0.81, 1.18]	0.6869
PLR (median [IQR])	147.23 [117.02, 192.14]	150.64 [117.02, 193.35]	140.98 [117.06, 180.65]	0.4168
NLR (median [IQR])	2.20 [1.70, 3.01]	2.13 [1.70, 3.04]	2.25 [1.70, 2.99]	0.5441
LMR (median [IQR])	4.15 [3.16, 5.28]	4.30 [3.26, 5.49]	4.06 [3.11, 5.20]	0.2885

Abbreviations: RT, radiotherapy; CCRT, concurrent chemoradiotherapy; IC, induction chemotherapy; AC, adjuvant chemotherapy; ALB, albumin; BMI, body mass index; HGB, hemoglobin; LDH, lactate dehydrogenase; TC, total cholesterol; TG, triglycerides; HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; Apo A1, apolipoprotein A1; Apo B, apolipoprotein B; PLR, platelet-to-lymphocyte ratio; NLR, neutrophil-to-lymphocyte ratio; LMR, lymphocyte-monocyte ratio; IQR, interquartile range.

and PFS to plot a calibration curve to evaluate the accuracy of the nomogram. The concordance index (C-index) and receiver operating characteristic (ROC) curve were performed to compare the predictive ability of the 8th edition TNM staging and nomogram. The clinical effectiveness of prediction model was estimated by decision curve analysis (DCA). In the training cohort, the scores for each prognostic factor were summed to obtain the total risk scores for our model to determine the optimal cut-off value via "survminer", which was applied to classify all patients into two different risk groups. The Kaplan–Meier curves were plotted for two groups by the nomogram stratification and differences in survival were compared using Log rank test. All related

Variables	OS		PFS	
	HR (95% CI)	P-value	HR (95% CI)	P-value
Age (years)				
≤61	Reference		Reference	
>61	2.15 (1.27–3.65)	0.004	1.55 (0.93–2.60)	0.095
Gender				
Female	Reference		Reference	
Male	1.64 (0.98–2.75)	0.060	1.32 (0.85–2.04)	0.215
Smoking				
No	Reference		Reference	
Yes	1.45 (0.96–2.20)	0.076	1.17 (0.81–1.71)	0.402
BMI (kg/m ²)				
≤22.65	Reference		Reference	
>22.65	0.72 (0.48–1.1)	0.131	0.78 (0.53–1.12)	0.180
T stage				
TI/T2	Reference		Reference	
T3/T4	1.85 (1.17–2.93)	0.009	1.75 (1.17–2.63)	0.006
N stage				
N0/N1	Reference		Reference	
N2/N3	2.04 (1.35–3.09)	0.001	2.10 (1.45–3.04)	<0.001
EBV DNA (co	pies/mL)			
≤7960	Reference		Reference	
>7960	2.01 (1.31–3.06)	0.001	2.00 (1.37–2.93)	<0.001
HGB (g/L)				
≤121	Reference		Reference	
>121	1.62 (0.84–3.13)	0.148	1.03 (0.62–1.70)	0.916
ALB (g/L)				
≤38.6	Reference		Reference	
>38.6	0.59 (0.35–0.98)	0.040	0.68 (0.42-1.08)	0.102

Table 2 Summarization of Univariate Analysis for 613 Cases Non-Metastatic NPC

(Continued)

Table 2 (Continued).

Variables	OS		PFS	
	HR (95% CI)	P-value	HR (95% CI)	P-value
LDH (U/L)			· · · ·	
≤195	Reference		Reference	
>195	2.10 (1.39–3.18)	<0.001	2.05 (1.42–2.97)	<0.001
TC (mmol/L)			· · · ·	
≤5.17	Reference		Reference	
>5.17	0.66 (0.43–1.03)	0.065	0.69 (0.47–1.01)	0.057
TG (mmol/L)			· · · ·	
≤1.56	Reference		Reference	
>1.56	1.53 (1.02–2.31)	0.042	1.25 (0.86–1.81)	0.241
HDL-C (mmo	I/L)		· · · · · ·	
≤0.98	Reference		Reference	
>0.98	0.41 (0.25–0.69)	0.001	0.49 (0.30–0.79)	0.003
LDL-C (mmol	/L)			
≤4.09	Reference		Reference	
>4.09	0.61 (0.32–1.14)	0.123	0.78 (0.47–1.31)	0.349
Apo Al (g/L)				
≤1.07	Reference		Reference	
>1.07	0.69 (0.43–1.11)	0.126	0.68 (0.44–1.03)	0.069
Apo B (g/L)				
≤1.18	Reference		Reference	
>1.18	0.68 (0.38–1.23)	0.204	0.89 (0.55–1.44)	0.632
PLR				
≤165.00	Reference		Reference	
>165.00	1.35 (0.90-2.04)	0.148	1.33 (0.92–1.91)	0.132
NLR				
≤2.85	Reference		Reference	
>2.85	1.81 (1.19–2.75)	0.005	1.64 (1.12–2.39)	0.011
LMR				
≤2.48	Reference		Reference	
>2.48	0.47 (0.28–0.79)	0.005	0.48 (0.30-0.78)	0.003

Abbreviations: BMI, body mass index; HGB, hemoglobin; ALB, albumin; LDH, lactate dehydrogenase; TC, total cholesterol; TG, triglycerides; HDL-C, high-density lipoprotein cholesterol; LDL-C, low-density lipoprotein cholesterol; Apo A1, apolipoprotein A1; Apo B, apolipoprotein B; PLR, platelet-to-lymphocyte ratio; NLR, neutrophil-to-lymphocyte ratio; LMR, lymphocyte–monocyte ratio.

statistical analysis and model building were performed in R software (Version 4.1.2) and SPSS (version 25.0). Statistical significance was considered by P < 0.05 for all results.

Results

Participant Characteristics

The clinical characteristics are described in Table 1 for all eligible non-metastatic NPC patients, without statistically significant differences found between the training and validation cohorts (p = 0.1090-0.9776). Of the 613 patients, the mean age at pathologically confirmed non-metastatic NPC diagnosis was 48 years (range 16-78 years), 442 (72.10%) were male, and 171 (27.90%) were female. Most patients were treated with CCRT (300 [48.94%]) or IC + CCRT (203 [33.12%]), while a few patients received RT alone (66 [10.77%]) or CCRT + AC (44 [7.18%]). The patients were classified into the following categories based on the TNM staging: stage I were 13 (2.12%), stage II were 140 (22.84%), stage III were 258 (42.09%), stage IVa were 202 (32.95%). The median follow-up time for the study population was 67 months (range 2–141 months). In the whole cohort, the 3- and 5-year OS rates were 88.6% and 81.9%, respectively, while the 3- and 5-year PFS rates were 80.7% and 74.6%, respectively. In the training cohort, the 3- and 5-year OS rates were 87.7% and 81.4%, respectively, while the 3- and 5-year PFS rates were 81.5% and 75.7%, respectively. In the validation cohort, the 3- and 5-year OS rates were 90.6% and 83.0%, respectively, while the 3- and 5-year PFS rates were 79.0% and 72.2%, respectively. The median values of HGB, LDH, TC, TG, HDL-C, LDL-C, Apo A1, Apo B, PLR, NLR and LMR were 140.00 g/L (interquartile range [IQR] 126.00-149.00 g/L), 176.00 g/L (IQR 153.00-203.00 g/L), 5.01 mmol/L (IQR 4.38-5.79 mmol/L), 1.25 mmol/L (IQR 0.93-1.90 mmol/L), 1.26 mmol/L (IQR 1.12-1.44 mmol/L), 3.27 mmol/L (IQR 2.74-3.93 mmol/L), 1.24g/L (IQR 1.09-1.42 g/L), 0.97 g/L (IQR 0.82-1.14 g/L), 147.23 (IQR 117.02-192.14), 2.20 (IQR 1.70-3.01) and 4.15 (IQR 3.16-5.28), respectively.

Variables	OS		PFS	
	HR (95% CI)	P-value	HR (95% CI)	P-value
Age (years)			Not selected	NA
≤61	Reference			
>61	2.20 (1.25–3.85)	0.006		
T stage				
TI/T2	Reference		Reference	
T3/T4	1.82 (1.13–2.92)	0.013	1.60 (1.06–2.41)	0.026
N stage			· · ·	
N0/N1	Reference		Reference	
N2/N3	1.54 (0.98–2.42)	0.062	1.74 (1.18–2.56)	0.005
EBV DNA (copies	;/mL)		· · ·	
≤7960	Reference		Reference	
>7960	1.89 (1.19–3.00)	0.007	1.80 (1.21–2.68)	0.004
ALB (g/L)			Not selected	NA
≤38.6	Reference			
>38.6	0.66 (0.38–1.13)	0.132		

Table 3 Summarization of Multivariate Analysis for 613 Cases Non-Metastatic NPC

(Continued)

Variables	OS		PFS	
	HR (95% CI)	P-value	HR (95% CI)	P-value
LDH (U/L)				
≤195	Reference		Reference	
>195	1.67 (1.08–2.59)	0.021	1.66 (1.13–2.44)	0.010
TG (mmol/L)			Not selected	NA
≤1.56	Reference			
>1.56	1.56 (1.00–2.43)	0.0496		
HDL-C (mmol/L)				
≤0.98	Reference		Reference	
>0.98	0.42 (0.24–0.73)	0.002	0.43 (0.26–0.71)	0.001
NLR				
≤2.85	Reference		Reference	
>2.85	1.31 (0.82–2.08)	0.263	1.12 (0.73–1.72)	0.589
LMR				
≤2.48	Reference		Reference	
>2.48	0.87 (0.48–1.60)	0.658	0.72 (0.42–1.24)	0.235

Table 3 (Continued).

Abbreviations: ALB, albumin; LDH, lactate dehydrogenase; TG, triglycerides; HDL-C, high-density lipoprotein cholesterol; NLR, neutrophil-to-lymphocyte ratio; LMR, lymphocyte-monocyte ratio.

Independent Risk Factors

The optimal cut-off values for age, BMI, EBV DNA, HGB, ALB, TC, TG, HDL-C, LDL-C, Apo A1, Apo B, LDH, PLR, NLR and LMR determined by R package "survminer" were 61 years, 22.65 kg/m2, 7960 copies/mL, 121 g/L, 38.6 g/L, 5.17 mmol/L, 1.56 mmol/L, 0.98 mmol/L, 4.09 mmol/L, 1.07 g/L, 1.18 g/L, 195 U/L, 165.00, 2.85 and 2.48, respectively. Univariate analysis demonstrated statistically significant risk factors with poorer OS in the training cohort including age

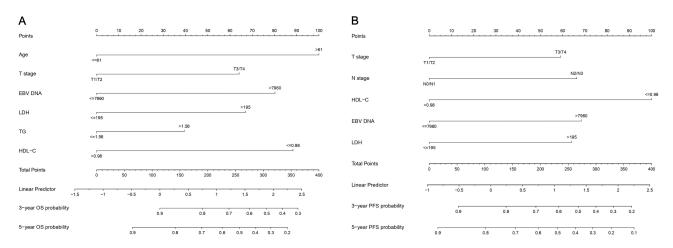


Figure I OS nomogram (A), including age, T stage, EBV DNA, TG, HDL-C and LDH. PFS nomogram (B), including T stage, N stage, EBV DNA, HDL-C and LDH. Abbreviations: TG, triglycerides; HDL-C, high-density lipoprotein cholesterol; LDH, lactate dehydrogenase; OS, overall survival; PFS, progression-free survival. >61 years (P = 0.004), T stage = T3 or T4 (P = 0.009), N stage = N2 or N3 (P = 0.001), EBV DNA >7960 (P = 0.001), ALB \leq 38.6 (P = 0.040), LDH >195 (P < 0.001), TG >1.56 (P = 0.042), HDL-C \leq 0.98 (P = 0.001), NLR >2.85 (P = 0.005) and LMR \leq 2.48 (P = 0.005). Univariate analysis demonstrated statistically significant risk factors with worse PFS in the training cohort including T stage = T3 or T4 (P = 0.006), N stage = N2 or N3 (P < 0.001), EBV DNA >7960 (P < 0.001), LDH >195 (P < 0.001), HDL-C \leq 0.98 (P = 0.003), NLR >2.85 (P = 0.011) and LMR \leq 2.48 (P = 0.003) (Table 2). The multivariate analysis identified that independent predictors were related to OS with P < 0.05 based on univariate analysis as follows: age, T stage, EBV DNA, LDH, TG and HDL-C. The multivariate analysis identified that independent predictors were related to PFS with P < 0.05 based on univariate analysis as follows: T stage, N stage, LDH, EBV DNA and HDL-C (Table 3).

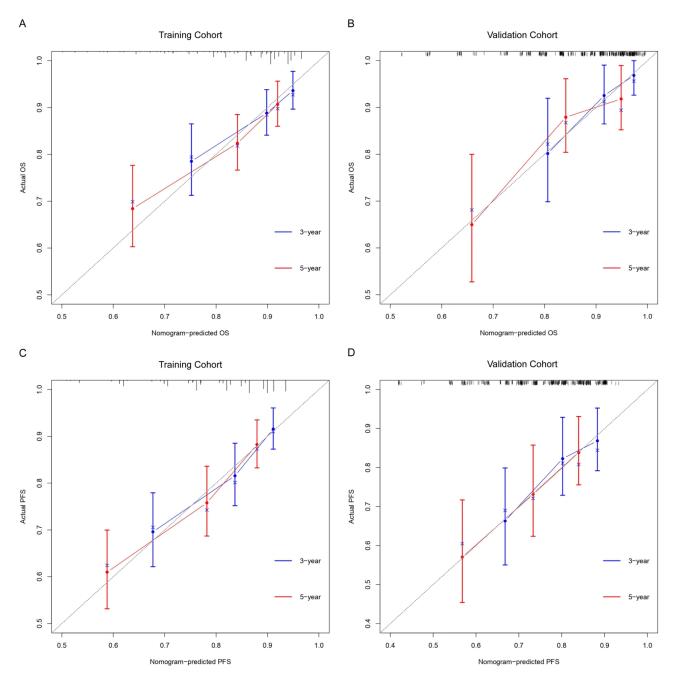


Figure 2 Calibration curves for predicting OS (A and B) and PFS (C and D) in the training and validation cohorts. Abbreviations: OS, overall survival; PFS, progression-free survival.

Nomogram Construction and Validation

As displayed in Figure 1, the nomogram was constructed for predicting OS and PFS at 3- and 5-year, including the independent predictors selected in the multivariate analysis. OS nomogram indicated that age and HDL-C had a more significant effect on OS, and PFS nomogram indicated that HDL-C had the most excellent effect on PFS. It can be seen from the calibration curves of OS and PFS that the nomogram had remarkable correspondence between the prediction outcomes and the actual observations in the two cohorts (Figure 2). According to the results presented in Table 4, the C-indexes for our proposed nomogram to predict OS and PFS were over 0.650, which was considerably greater than the 8th TNM staging in the training and validation cohorts. Similarly, the ROC curves demonstrated that the nomogram had greater AUC areas and more satisfactory discrimination capability than the current TNM staging, both in the training and validation cohorts (Figure 3). Furthermore, the DCA curves revealed that the nomogram had a broader range of threshold probabilities to predict OS and PFS compared with TNM staging, resulting in higher net clinical benefits (Figure 4).

Nomograms for Risk Stratification

206 and 125 were the respective cut-off values for OS and PFS nomograms, obtained from the total scores for the constructed nomogram. For estimating the risk stratification capability of nomogram, all patients were classified as a low-(total scores of OS: \leq 206; total scores of PFS: \leq 125) and high-risk group (total scores of OS: \geq 206; total scores of PFS: \geq 125). As clearly seen in the Kaplan–Meier curves (Figure 5), patients with low risk revealed better-predicted outcomes than those with high risk in the training cohort. Different risk groups showed considerable differences (all P < 0.05), which was also verified in the validation cohort.

Discussion

From what we know, this is the first research to characterize the establishment and verification of a nomogram incorporating lipid metabolism markers and inflammatory biomarker to predict OS and PFS for non-metastatic NPC. Our nomogram demonstrated superior predictive power over the 8th edition TNM staging, which would help clinicians determine whether patients are at high risk for poorer prognosis and develop individualized therapeutic regimens.

Currently, the 8th edition TNM staging is the most extensively acknowledged gold standard to predict prognosis and guide decision-making on treatment in clinical practice. However, there is still a significant difference in the prognosis among patients with identical staging, even after receiving comparable therapy.²⁷ The comprehensive prediction capability of the TNM staging system is insufficient without considering the importance of other risk factors, resulting in this phenomenon. Thus, we explore a reliable tool to complement the TNM staging that accurately recognizes the potential patients at high risk of needing intensive treatment to obtain more clinical benefits.

Hematologic markers are economical and easy to detect to act as potential prognostic variables for NPC. Previous research shows that plasma EBV DNA plays a well-defined role in predicting and monitoring the possibility of recurrence, metastasis and

Variables	Training Cohort	Validation Cohort
	C-Index (95% CI)	C-Index (95% CI)
os		
Nomogram	0.687 (0.630–0.744)	0.717 (0.630–0.804)
TNM stage	0.570 (0.534–0.606)	0.614 (0.556–0.672)
PFS		
Nomogram	0.670 (0.621–0.719)	0.651 (0.574–0.728)
TNM stage	0.570 (0.538–0.602)	0.572 (0.515–0.629)

Table 4 C-Indexes of Nomogram Compared with TNM Staging

Abbreviations: OS, overall survival; PFS, progression-free survival.

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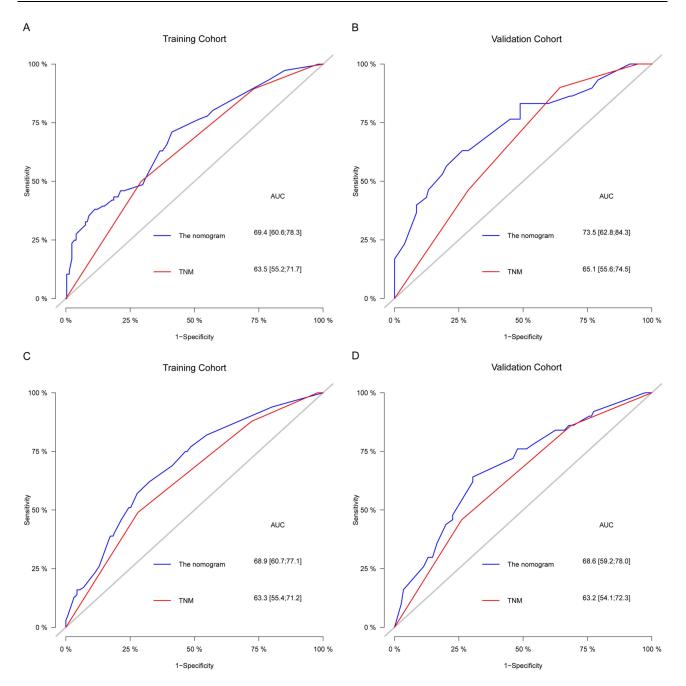


Figure 3 ROC curves to predict OS (A and B) and PFS (C and D) at 5 years to compare with TNM staging.

death during the whole treatment period of NPC.^{8,28,29} It can reflect tumor load and serve as a biomarker to complement TNM staging and improve the prognosis of patients.^{30,31} Similarly, our results supported the favorable effect of plasma EBV DNA in prediction. In addition, our study combined indicators reflecting lipid metabolism and inflammation status of patients in the same prognostic model. A large amount of evidence has shown that tumors satisfy the need to grow rapidly and progress aggressively by regulating lipid metabolism during development.^{13–15,32} TG, as one of the energy sources for tumor cells, enhances tumor cell proliferation and growth under hypoxic and hypoperfusion environment, as well as exerts protective effects.^{33,34} There is a close correlation between a poor prognosis and high TG levels in malignancies including breast cancer, prostate cancer and non-small cell lung cancer.^{20,35,36} Increased TG was observed to be beneficial for adverse prognostic outcome occurrences in NPC patients by Peng et al.²¹

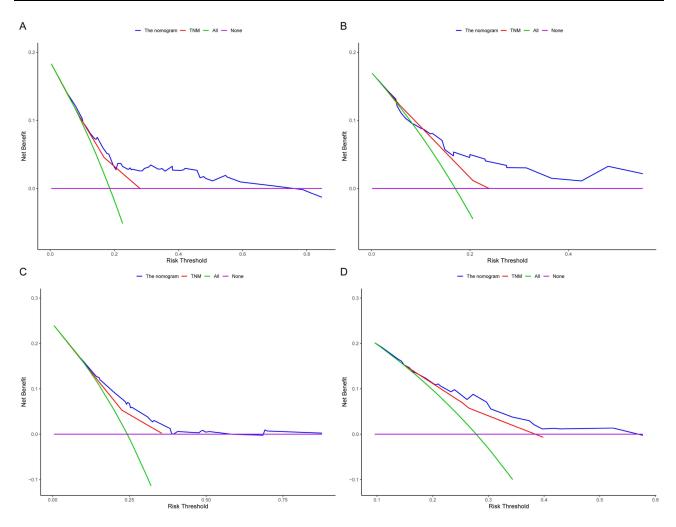


Figure 4 Decision curve analysis of 5 years OS (A and B) and PFS (C and D) in the training and validation cohorts.

Observations to date have proven that systemic inflammation is involved in the occurrence and development of malignancies.¹⁶ As is well known, several properties of HDL have been discovered to be anti-inflammatory, anti-oxidative and anti-atherosclerotic.³⁷ Under the stimulation of inflammatory environment, HDL is responsible for inhibiting inflammationdriven tumor angiogenesis by exerting its anti-inflammatory properties.³⁸ Meanwhile, HDL participates in regulating the cholesterol efflux mechanism to maintain the balance of cholesterol that contributes to its anti-tumor role.³⁹ Li et al⁴⁰ illustrated that high HDL-C levels were significantly correlated with improved OS for NPC patients. As shown in the study by Wang et al,²³ HDL-C was a reliable prognostic indicator for risk stratification of NPC. Our findings demonstrated that TG and HDL-C were independent predictors for NPC after treatment with IMRT, which is consistent with previous studies. A number of inflammatory biomarkers were examined considering the importance of inflammation in tumor development, including LDH, PLR, NLR and LMR. LDH was the only independent risk prognostic factor identified by multivariate analysis. Wan et al²⁴ analyzed that LDH was an adverse prognostic indicator for NPC reflecting inflammatory status. LDH is a key enzyme for maintaining cancer cell survival in a hypoxic environment by facilitating the conversion of pyruvate to lactate.⁴¹

Recently, molecular markers such as long noncoding RNA⁴² and microRNAs⁴³ have been confirmed to be significant in predicting the prognosis of NPC. However, molecular markers were not included in the nomogram because they were expensive and not routinely tested in our hospital. In contrast, three advantages can be found in the predictive model we established: first, lipid metabolism markers are more economical, objective, and easily measurable compared to plasma EBV DNA detection, and they are a common test for hospital admission. Second, the predictive and integration capabilities of nomogram are superior to traditional TNM staging, allowing for more accurate risk

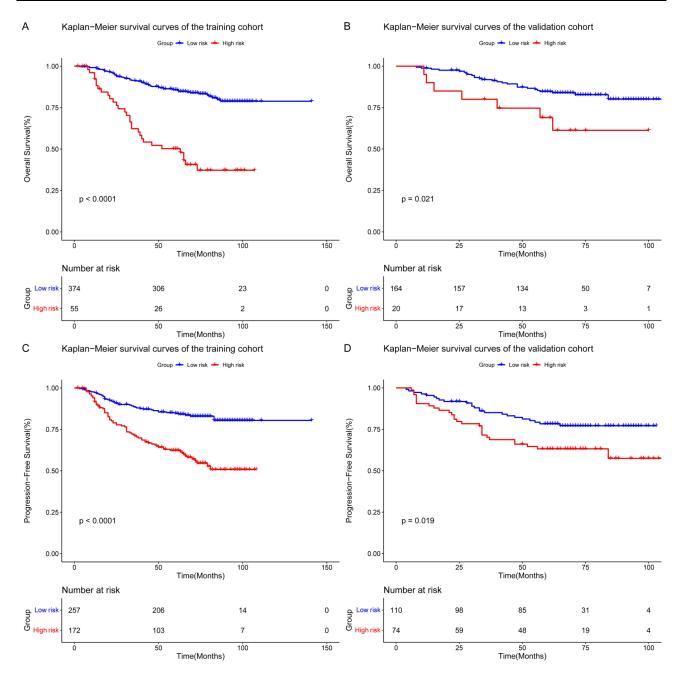


Figure 5 Kaplan-Meier plots for two different risk groups according to OS (A and B) and PFS (C and D) nomogram in the training and validation cohorts.

stratification and intensive therapies. Third, all of the prognostic factors contained in the nomogram are readily available, enabling them to be widely used even in primary medical institutions, providing practical assistance to clinicians in their medical work.

To be honest, our study has some limitations that cannot be neglected. First, the sample size of our study needs to be larger, and it is only sourced from areas where NPC is endemic, which may not be applicable for generalizing to other non-endemic areas. Second, the data for all patients are from the same hospital, and there needs to be more data from multiple centers to perform external validation of our results. Third, no authoritative, unified standards for the detection of plasma EVB DNA have been published.

Conclusion

In summary, our proposed nomogram containing TG, HDL-C and LDH was established and verified for predicting OS and PFS in non-metastatic NPC patients. Compared to the TNM staging, our nomogram can determine whether patients are at high risk for poorer prognosis accurately and early, which provides a strong foundation for clinicians to select appropriate intensive treatment. The combination of lipid metabolic markers and inflammatory index in nomogram is economical, objective, and easily measurable, facilitating its widespread use in clinical practice.

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Disclosure

The authors report no conflicts of interest in this work.

References

- 1. Chen YP, Chan ATC, Le QT, Blanchard P, Sun Y, Ma J. Nasopharyngeal carcinoma. Lancet. 2019;394(10192):64-80. doi:10.1016/S0140-6736(19) 30956-0
- Sung H, Ferlay J, Siegel RL, et al. Global cancer statistics 2020: GLOBOCAN estimates of Incidence and mortality worldwide for 36 cancers in 185 countries. CA Cancer J Clin. 2021;71(3):209–249. doi:10.3322/caac.21660
- 3. Blanchard P, Lee A, Marguet S, et al. Chemotherapy and radiotherapy in nasopharyngeal carcinoma: an update of the MAC-NPC meta-analysis. *Lancet Oncol.* 2015;16(6):645–655. doi:10.1016/S1470-2045(15)70126-9
- 4. Sun XS, Liu SL, Luo MJ, et al. The association between the development of radiation therapy, image technology, and chemotherapy, and the survival of patients with nasopharyngeal carcinoma: a cohort study from 1990 to 2012. *Int J Radiat Oncol.* 2019;105(3):581–590. doi:10.1016/j. ijrobp.2019.06.2549
- Zhang MX, Li J, Shen GP, et al. Intensity-modulated radiotherapy prolongs the survival of patients with nasopharyngeal carcinoma compared with conventional two-dimensional radiotherapy: a 10-year experience with a large cohort and long follow-up. *Eur J Cancer.* 2015;51(17):2587–2595. doi:10.1016/j.ejca.2015.08.006
- 6. Lee AWM, Ng WT, Chan LLK, et al. Evolution of treatment for nasopharyngeal cancer success and setback in the intensity-modulated radiotherapy era. *Radiother Oncol.* 2014;110(3):377–384. doi:10.1016/j.radonc.2014.02.003
- 7. Lee AWM, Ng WT, Chan ATC. Management of nasopharyngeal carcinoma: current practice and future perspective. J Clin Oncol. 2015;33 (29):3356–3364. doi:10.1200/JCO.2015.60.9347
- 8. Lam WKJ, Chan KCA. Plasma Epstein-Barr virus DNA as an archetypal circulating tumour DNA marker. J Pathol. 2019;247(5):641-649. doi:10.1002/path.5249
- 9. Le QT, Zhang Q, Cao H, et al. An international collaboration to harmonize the quantitative plasma Epstein-Barr Virus DNA assay for future biomarker-guided trials in nasopharyngeal carcinoma. *Clin Cancer Res.* 2013;19(8):2208–2215. doi:10.1158/1078-0432.CCR-12-3702
- 10. Luo W. Nasopharyngeal carcinoma ecology theory: cancer as multidimensional spatiotemporal "unity of ecology and evolution" pathological ecosystem. *Theranostics*. 2023;13(5):1607–1631. doi:10.7150/thno.82690
- 11. Muluh TA, Sheng SX, Ying Y. Targeting cancer metabolic vulnerabilities for advanced therapeutic efficacy. *Biomed Pharmacother*. 2023;162:114658. doi:10.1016/j.biopha.2023.114658
- 12. Corn KC, Windham MA, Rafat M. Lipids in the tumor microenvironment: from cancer progression to treatment. *Prog Lipid Res.* 2020;80:101055. doi:10.1016/j.plipres.2020.101055
- 13. Bian X, Liu R, Meng Y, Xing D, Xu D, Lu Z. Lipid metabolism and cancer. J Exp Med. 2021;218(1):e20201606. doi:10.1084/jem.20201606
- 14. Luo X, Cheng C, Tan Z, et al. Emerging roles of lipid metabolism in cancer metastasis. *Mol Cancer*. 2017;16(1):76. doi:10.1186/s12943-017-0646-3 15. Cheng C, Geng F, Cheng X, Guo D. Lipid metabolism reprogramming and its potential targets in cancer. *Cancer Commun*. 2018;38(1):27.
- doi:10.1186/s40880-018-0301-4 16. Coussens LM, Werb Z. Inflammation and cancer. *Nature*. 2002;420(6917):860–867. doi:10.1038/nature01322
- His M, Zelek L, Deschasaux M, et al. Prospective associations between serum biomarkers of lipid metabolism and overall, breast and prostate cancer risk. *Eur J Epidemiol.* 2014;29(2):119–132. doi:10.1007/s10654-014-9884-5
- 18. Yang MH, Rampal S, Sung J, et al. The association of serum lipids with colorectal adenomas. Am J Gastroenterol. 2013;108(5):833-841. doi:10.1038/ajg.2013.64
- 19. Lyu J, Yang N, Guan W, et al. Post-treatment serum triglyceride: an effective biomarker for body fat mass and overall survival in esophageal squamous cell cancer patients treated with chemoradiotherapy. *Front Nutr.* 2022;9:1050643. doi:10.3389/fnut.2022.1050643

- Ma C, Wang X, Guo J, Liu P. Prognostic significance of preoperative serum triglycerides and high-density lipoproteins cholesterol in patients with non-small cell lung cancer: a retrospective study. *Lipids Health Dis.* 2021;20(1):69. doi:10.1186/s12944-021-01492-y
- 21. Peng H, Chen L, Zhang Y, et al. Prognostic value of diabetes in patients with nasopharyngeal carcinoma treated with intensity-modulated radiation Therapy. Sci Rep. 2016;6(1):22200. doi:10.1038/srep22200
- 22. Xie Z, Shao Y. The predictive value of serum lipids for eye metastases in male nasopharyngeal carcinoma patients. *Biosci Rep.* 2020;40(6): BSR20201082. doi:10.1042/BSR20201082
- Wang CT, Chen MY, Guo X, et al. Association between pretreatment serum high-density lipoprotein cholesterol and treatment outcomes in patients with locoregionally advanced nasopharyngeal carcinoma treated with chemoradiotherapy: findings from a randomised trial. J Cancer. 2019;10 (16):3618–3623. doi:10.7150/jca.32621
- 24. Wan XB, Wei L, Li H, et al. High pretreatment serum lactate dehydrogenase level correlates with disease relapse and predicts an inferior outcome in locally advanced nasopharyngeal carcinoma. *Eur J Cancer*. 2013;49(10):2356–2364. doi:10.1016/j.ejca.2013.03.008
- 25. Li XH, Chang H, Xu BQ, et al. An inflammatory biomarker-based nomogram to predict prognosis of patients with nasopharyngeal carcinoma: an analysis of a prospective study. *Cancer Med*. 2017;6(1):310–319. doi:10.1002/cam4.947
- 26. Zhao R, Liang Z, Chen K, Zhu X. Nomogram based on inflammatory biomarkers and nutritional indicators for predicting overall survival in locoregionally advanced nasopharyngeal carcinoma. *J Inflamm Res.* 2022;15:2971–2981. doi:10.2147/JIR.S366299
- Guo R, Wang S, Hu Y, et al. New parameters of the 8th edition AJCC/UICC T category in nasopharyngeal carcinoma: cervical vertebrae invasion and parotid gland invasion. *Clin Transl Med.* 2020;10(7). doi:10.1002/ctm2.202
- Hui EP, Li WF, Ma BB, et al. Integrating postradiotherapy plasma Epstein–Barr virus DNA and TNM stage for risk stratification of nasopharyngeal carcinoma to adjuvant therapy. Ann Oncol. 2020;31(6):769–779. doi:10.1016/j.annonc.2020.03.289
- 29. Gihbid A, Benzeid R, Faouzi A, et al. The dynamic change in plasma Epstein-Barr Virus DNA load over a long-term follow-up period predicts prognosis in nasopharyngeal carcinoma. Viruses. 2023;15:66.
- 30. Tan R, Phua SKA, Soong YL, et al. Clinical utility of Epstein-Barr virus DNA and other liquid biopsy markers in nasopharyngeal carcinoma. *Cancer Commun.* 2020;40(11):564–585. doi:10.1002/cac2.12100
- 31. Guo R, Tang LL, Mao YP, et al. Proposed modifications and incorporation of plasma Epstein-Barr virus DNA improve the TNM staging system for Epstein-Barr virus-related nasopharyngeal carcinoma: refined TNM Staging for EBV-Related NPC. Cancer. 2019;125(1):79–89. doi:10.1002/ cncr.31741
- 32. Cao Y. Adipocyte and lipid metabolism in cancer drug resistance. J Clin Invest. 2019;129(8):65.
- 33. Ackerman D, Tumanov S, Qiu B, et al. Triglycerides promote lipid homeostasis during hypoxic stress by balancing fatty acid saturation. *Cell Rep.* 2018;24(10):2596–2605.e5. doi:10.1016/j.celrep.2018.08.015
- 34. Snaebjornsson MT, Janaki-Raman S, Schulze A. Greasing the wheels of the cancer Machine: the role of lipid metabolism in cancer. *Cell Metab.* 2020;31(1):62–76. doi:10.1016/j.cmet.2019.11.010
- 35. Berrino F, Villarini A, Traina A, et al. Metabolic syndrome and breast cancer prognosis. Breast Cancer Res Treat. 2014;147(1):159-165. doi:10.1007/s10549-014-3076-6
- 36. Allott EH, Howard LE, Cooperberg MR, et al. Serum lipid profile and risk of prostate cancer recurrence: results from the SEARCH Database. *Cancer Epidemiol Biomarkers Prev.* 2014;23(11):2349–2356. doi:10.1158/1055-9965.EPI-14-0458
- Tan JTM, Bursill CA. The role of high-density lipoproteins in the regulation of angiogenesis. Cardiovasc Res. 2015;106(2):184–193. doi:10.1093/ cvr/cvv104
- 38. Zhao T, Zhu N, Shi Y, et al. Targeting HDL in tumor microenvironment: new hope for cancer therapy. J Cell Physiol. 2021;236(11):7853–7873. doi:10.1002/jcp.30412
- 39. Ganjali S, Ricciuti B, Pirro M, et al. High-Density lipoprotein components and functionality in cancer: state-of-the-Art. *Trends in Endocrinol Metab.* 2019;30(1):12–24. doi:10.1016/j.tem.2018.10.004
- 40. Li J, Chen S, Peng S, et al. Prognostic nomogram for patients with nasopharyngeal carcinoma incorporating hematological biomarkers and clinical characteristics. *Int J Biol Sci.* 2018;14.
- Forkasiewicz A, Dorociak M, Stach K, Szelachowski P, Tabola R, Augoff K. The usefulness of lactate dehydrogenase measurements in current oncological practice. *Cell Mol Biol Lett.* 2020;25(1):35. doi:10.1186/s11658-020-00228-7
- 42. Liang YL, Zhang Y, Tan XR, et al. A IncRNA signature associated with tumor immune heterogeneity predicts distant metastasis in locoregionally advanced nasopharyngeal carcinoma. *Nat Commun.* 2022;13(1):2996. doi:10.1038/s41467-022-30709-6
- 43. Yang S, Li Y. MicroRNAs: novel factors in clinical diagnosis and prognosis for nasopharyngeal carcinoma. *Acta Pharmacol Sin.* 2012;33 (8):981–982. doi:10.1038/aps.2012.98

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