

Establishment and validation of nomograms for predicting survival of lung invasive adenocarcinoma based on the level of pathological differentiation: a SEER cohort-based analysis

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Background: The pathological differentiation of invasive adenocarcinoma (IAC) has been linked closely with epidemiological characteristics and clinical prognosis. However, the current models cannot accurately predict IAC outcomes and the role of pathological differentiation is confused. This study aimed to establish differentiation-specific nomograms to explore the effect of IAC pathological differentiation on overall survival (OS) and cancer-specific survival (CSS).

Methods: The data of eligible IAC patients between 1975 and 2019 were collected from the Surveillance, Epidemiology, and End Results (SEER) database, and randomly divided in a ratio of 7:3 into a training cohort and a validation cohort. The associations between pathological differentiation and other clinical characteristics were evaluated using chi-squared test. The OS and CSS analyses were performed using the Kaplan-Meier estimator, and the log-rank test was used for nonparametric group comparisons. Multivariate survival analysis was performed using a Cox proportional hazards regression model. The discrimination, calibration, and clinical performance of nomograms were assessed by area under receiver operating characteristic curve (AUC), calibration plots, and decision curve analysis (DCA).

Results: A total of 4,418 IAC patients (1,001 high-differentiation, 1,866 moderate-differentiation, and 1,551 low-differentiation) were identified. Seven risk factors [age, sex, race, tumor-node-metastasis (TNM) stage, tumor size, marital status, and surgery] were screened to construct differentiation-specific nomograms. Subgroup analyses showed that disparate pathological differentiation played distinct roles in prognosis, especially in patients with older age, white race, and higher TNM stage. The AUC of nomograms for OS and CSS in the training cohort were 0.817 and 0.835, while in the validation cohort were 0.784 and 0.813. The calibration curves showed good conformity between the prediction of the nomograms and the actual observations. DCA results indicated that these nomogram models could be used as a supplement to the prediction of the TNM stage.

Conclusions: Pathological differentiation should be considered as an independent risk factor for OS and CSS of IAC. Differentiation-specific nomogram models with good discrimination and calibration capacity were developed in the study to predict the OS and CSS in 1-, 3- and 5-year, which could be used predict prognosis and select appropriate treatment options.

Keywords: Surveillance, Epidemiology, and End Results (SEER); invasive adenocarcinoma (IAC); nomogram model; prognosis

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Introduction

Lung cancer remains the main reason for cancer-related death on a global scale and lung adenocarcinoma (LUAD) is the most common pathological type (1). To ameliorate this major global health problem, further research is needed to focus on the prognosis. Many reports were published on prognostic risk factors for LUAD: factors like age, sex, race, tumor-node-metastasis (TNM) stage, pathological grade, marital status, and therapeutic regimen are closely related to survival. The age of the patients is a confirmed risk factor for LUAD, and it always affects the type of therapeutic strategy (2,3). The majority of studies consider females are more likely to have a better prognosis, which may relate to the estrogen receptor beta (ER- β) pathway (2,4-7). The race is also considered to be an independent prognostic factor for LUAD, with Asians, Pacific Islanders, and Hispanics experiencing better survival (8,9). In tons of former studies, TNM stage, pathological grade, and therapeutic regimen are used as prognostic factors of LUAD, thus almost all lung cancer nomograms include these three factors (3,7,10-14). There are 4 histologic subtypes of LUAD according to the

Highlight box

Key findings

• This study analyzed clinicopathological features of invasive adenocarcinoma (IAC) and investigated the risk factors associated with prognosis. Nomogram models of different pathological stages of IAC were established to predict prognosis.

What is known and what is new?

- Nomograms are widely used in various tumors, and prediction models for IAC are emerging in an endless stream.
- The relationship between pathological differentiation and other clinicopathologic factors was analyzed in this study. IAC and the variants were classified into three groups: high, medium, and low differentiation, and models of each group were constructed respectively.

What is the implication, and what should change now?

 This study highlights the importance of pathological differentiation in IAC survival and provides clinical nomograms for IAC patients with different pathologies. The pathogenesis and genomic changes of IAC with different histology might be further explored in the future. 2015 WHO classification: preinvasive lesions, minimally invasive adenocarcinoma (MIA), invasive adenocarcinoma (IAC), and adenocarcinoma variants (15,16). Compared with preinvasive lesions and MIA, five subtypes of IAC and its variants portend a significantly worse prognosis. But whether these prognostic factors affecting LUAD still have effects on IAC and its variants has not been confirmed.

The International Association for the Study of Lung Cancer (IASLC) pathology panel established a set of histologic standards to assess the outcomes of lung cancer. Invasive pulmonary adenocarcinoma was divided into three types according to the degree of differentiation: grade 1, lepidic predominant tumor only; grade 2, acinar or papillary predominant tumor; and grade 3, solid, micropapillary, or complex gland (including variants) (17). Motivated by this, the pathological types were classified into three broad categories based on the grade of differentiation: high differentiation (lepidic adenocarcinoma only; LEP), moderate differentiation (acinar adenocarcinoma or papillary adenocarcinoma; ACI/PAP) and low differentiation (micropapillary adenocarcinoma, solid adenocarcinoma or variants; MIP/SOL).

Contemporary therapeutic decisions for IAC patients are often based on the TNM stage, which does not account for different pathological subtypes. Several recent studies have constructed predicting nomogram models based on specific pathological subtypes such as invasive mucinous adenocarcinoma (18-20) and PAP (21). However, these models which focus on only one pathological type are not widely used. Some researchers have established a prognostic model totally based on pathologic features for IAC, which showed passable discriminative ability (22). The predictive accuracy of the models was satisfactory, nevertheless, they just considered the role of IAC pathology, which limited the scope of the model's application. There were also no established nomograms to estimate the outcome of IAC patients including multiple variants. So, it is warranted to analyze the predictive value of pathological differentiation combined with other clinical predictors.

This study was to analyze clinicopathological features of IAC and investigate the risk factors associated with prognosis. Furthermore, the relationship between pathological differentiation and other risk factors were 806

probed. Nomogram models were established to better predict IAC prognosis. We present the following article in accordance with the TRIPOD reporting checklist (available at https://tcr.amegroups.com/article/view/10.21037/tcr-22-2308/rc).

Methods

Study design

This study was to explore the mechanism by which the pathological differentiation affects IAC outcomes and establish differentiation-specific nomograms. By reviewing the data of the Surveillance, Epidemiology, and End Results (SEER) database, the overall characteristics of IAC patients could be obtained. All patients would be randomly divided in a ratio of 7:3 into a training cohort and a validation cohort. The training cohort was to screen risk factors of IAC and establish differentiation-specific nomograms. The training cohort was utilized for internal validation and the validation cohort was used for external validation. In view of the data collected from the SEER database, the follow-up procedures would be omitted. The methodology each step was described in detail in following subsections.

Data source and coborts selection

The patient data were gathered from the SEER public use database SEER 8 Regs Custom Data, Nov 2021 Sub (1975–2019), which collects clinicopathologic data of cancer patients, including demographics, tumor pathology, stage at diagnosis, treatment protocols, and survival from 1975 to 2019.

IAC patients between 1975 and 2019 were identified using SEER*Stat software (version 8.4.0). The inclusion criteria included two items: tumor primary sites and histologic ICD-O-3 codes. The primary sites were employed to identify the following sites: C34.0-Main bronchus; C34.1-Upper lobe, lung; C34.2-Middle lobe, lung; C34.3-Lower lobe, lung; C34.8-Overlapping lesion of lung; C34.9-Lung, NOS. The histologic ICD-O-3 codes (International Classification of Diseases for Oncology, Third Edition) were employed to identify the following subtypes: lepidic adenocarcinoma (ICD-O-3 code 8250), ACI (ICD-O-3 code 8551), PAP (ICD-O-3 code 8260), MIP (ICD-O-3 code 8265), SOL (ICD-O-3 code 8230), invasive mucinous adenocarcinoma (ICD-O-3 code 8253), mixed invasive mucinous and nonmucinous adenocarcinoma (ICD-O-3 code 8254), colloid adenocarcinoma (ICD-O-3 code 8480), fetal adenocarcinoma (ICD-O-3 code 8333), and enteric adenocarcinoma (ICD-O-3 code 8144).

The exclusion criteria were as follows: (I) no histological confirmation was found for the diagnosis; (II) the survival time was unknown or less than one month; (III) multiple primary cancers were diagnosed but IAC was not the first one; (IV) type of reporting source was autopsy or death certificate; (V) disease-related detail is missing, such as unknown age, gender, race, grade, TNM stage, and so on; (VI) therapeutic regimen is missing.

Final results revealed 4,418 patients with IAC met the criteria for inclusion and exclusion. For further analysis, 70% of screened patients were randomly divided into a training cohort and 30% were classified into a validation cohort to externally verify the final nomogram (*Figure 1*). Two cohorts showed no significant discrepancy in demographics and clinical characteristics by chi-squared test (*Table 1*).

Study variables

Several demographic variables of patients, including age, sex, race, and marital status, were extracted from the SEER database. For a more concise analysis, the patients were segmented into four 4 groups: <60 years old, 60–69 years old, 70–79 years old, and ≥80 years old. Marital status was categorized as married, single, separated, divorced, and widowed according to the database. Divorced and separated patients were assigned into the divorced group which makes the results clear and credible.

Clinicopathologic variables captured from the SEER database encompassed pathological subtypes, tumor staging according to the 8th American Joint Committee on Cancer (AJCC) TNM stage, and surgery. Pathological subtypes were classified into 3 groups according to differentiation depending on the previously mentioned criteria. High differentiation included only adenocarcinoma. Median differentiation consisted of acinar or PAP. Low differentiation included MIP, SOL, invasive mucinous adenocarcinoma, mixed invasive mucinous and nonmucinous adenocarcinoma, colloid adenocarcinoma, fetal adenocarcinoma, and enteric adenocarcinoma.

From diagnosis to death for any reason, overall survival (OS) was the primary terminal point. As a secondary terminal point, cancer-specific survival (CSS) was defined as the period from diagnosis to death due to IAC.



Figure 1 Flow diagram of the IAC patients with training and validation cohorts. IAC, invasive adenocarcinoma; AJCC, American Joint Committee on Cancer; TNM, tumor-node-metastasis; SEER, Surveillance, Epidemiology, and End Results.

Construction and verification of nomograms

Risk factors were identified by univariate analysis and multivariate analysis, and then the nomograms were constructed. The performance and accuracy of nomograms were assessed by receiver operating characteristics (ROC) curves, calibration plots, and decision curve analysis (DCA). More areas under the ROC curve (AUC) meant higher quality. Conventionally, 0.5 < AUC < 0.7 is considered low accuracy, 0.7 < AUC < 0.9 is medium accuracy, and AUC >0.9 is high accuracy. Most researchers believe that a qualified model should have an AUC greater than 0.7. Our calibrations were performed using bootstraps and 1,000 resamples to ensure a solid comparison of the predicted and observed OS and CSS over 1, 3, and 5 years. DCA curves showed benefits derived from constructed models.

Ethical statement

The study was conducted in accordance with the

Declaration of Helsinki (as revised in 2013). Since the present study relied on the SEER database, which was freely available to the public (http://seer.cancer.gov/seerstat/), the ethical approval was waived.

Statistical analysis

The consistency between the training cohort and the testing cohort was checked by chi-squared tests. By using this statistical method, the baseline data of demographical and clinicopathological features in disparate differentiation groups were compared. OS and CSS survival analyses were performed using the Kaplan-Meier method and nonparametric group comparisons were performed using the log-rank test. For the training cohort, the independent prognostic factors of OS and CSS were determined using univariate and multivariate Cox proportional hazards regression models. All P values were two-sided, and statistical significance was determined by P<0.05. All statistical analyses and drawings were conducted using R software version 4.2.1.

Results

Baseline patient features

Four thousand four hundred and eighteen eligible cases from 1975 to 2019 were screened through SEER database searches. Information of the patients is displayed in Tables 1,2. As shown in Table 1, 1,001 cases were highly differentiated (22.7%), 1,866 cases were moderately differentiated (42.2%), and 1,551 cases were low differentiated (35.1%). The factors, age, gender, race, marital status, stage, surgery, death, and cancer-specific death were shown to significantly influence on pathological differentiation (P<0.001, chi-square test). In view of the difference in data composition of disparate pathological groups, differentiation-specific nomograms were applied to predict the OS and CSS. In general, there were more seniors over the age of 60 (83.6%), females (59.9%), whites (80.8%), and married people (59.2%). Most patients were in the early stages (T1: 50.0%, N0: 80.5%, M0: 87.2%, TNM stage I: 61.1%) and had surgery (78.3%). The high differentiation group had the highest proportion of T1 (57.8%), N0 (87.5%), TNM stage I (68.2%), suggesting a potential connection between high-differentiation and early TNM stage. The median differentiation group had the highest percentage of surgery (87.1%). A comparison of

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Table 1 Demographics and clinicopathological characteristics by pathological differentiation subtypes of IAC

Characteristic	High (n=1,001)	Median (n=1,866)	Low (n=1,551)	Total (N=4,418)	P value
Age, years					<0.001
<60	128 (12.8)	313 (16.8)	284 (18.3)	725 (16.4)	
60–69	292 (29.2)	671 (36.0)	507 (32.7)	1,470 (33.3)	
70–79	395 (39.5)	675 (36.2)	540 (34.8)	1,610 (36.4)	
≥80	186 (18.6)	207 (11.1)	220 (14.2)	613 (13.9)	
Sex					<0.001
Female	667 (66.6)	1,140 (61.1)	841 (54.2)	2,648 (59.9)	
Male	334 (33.4)	726 (38.9)	710 (45.8)	1,770 (40.1)	
Race					<0.001
White	821 (82.0)	1,476 (79.1)	1,273 (82.1)	3,570 (80.8)	
Black	62 (6.2)	165 (8.8)	144 (9.3)	371 (8.4)	
Other	118 (11.8)	225 (12.1)	134 (8.6)	477 (10.8)	
Marital status					<0.001
Married	594 (59.3)	1,094 (58.6)	927 (59.8)	2,615 (59.2)	
Divorced	123 (12.3)	239 (12.8)	197 (12.7)	559 (12.6)	
Single	103 (10.3)	283 (15.2)	214 (13.8)	600 (13.6)	
Widowed	181 (18.1)	250 (13.4)	213 (13.7)	644 (14.6)	
Derived AJCC T stage					<0.001
T1	579 (57.8)	1,018 (54.6)	613 (39.5)	2,210 (50.0)	
T2	179 (17.9)	516 (27.7)	389 (25.1)	1,084 (24.4)	
Т3	41 (4.1)	156 (8.4)	148 (9.5)	345 (7.8)	
T4	202 (20.2)	176 (9.4)	401 (25.9)	779 (17.6)	
Derived AJCC N stage					<0.001
N0	876 (87.5)	1,513 (81.1)	1,168 (75.3)	3,557 (80.5)	
N1	46 (4.6)	131 (7.0)	141 (9.1)	318 (7.2)	
N2	68 (6.8)	177 (9.5)	188 (12.1)	433 (9.8)	
N3	11 (1.1)	45 (2.4)	54 (3.5)	110 (2.5)	
Derived AJCC M stage					<0.001
M0	887 (88.6)	1,693 (90.7)	1,272 (82.0)	3,852 (87.2)	
M1	114 (11.4)	173 (9.3)	279 (18.0)	566 (12.8)	
Derived AJCC stage					<0.001
I	683 (68.2)	1,250 (67.0)	765 (49.3)	2,698 (61.1)	
II	64 (6.4)	212 (11.4)	192 (12.4)	468 (10.6)	
III	140 (14.0)	231 (12.4)	315 (20.3)	686 (15.5)	
IV	114 (11.4)	173 (9.3)	279 (18.0)	566 (12.8)	

Table 1 (continued)

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Characteristic	High (n=1,001)	Median (n=1,866)	Low (n=1,551)	Total (N=4,418)	P value
Surgery					<0.001
Yes	689 (68.8)	1,626 (87.1)	1,143 (73.7)	3,458 (78.3)	
No	312 (31.2)	240 (12.9)	408 (26.3)	960 (21.7)	
Death					<0.001
Censored	602 (60.1)	1,606 (86.1)	941 (60.7)	3,149 (71.3)	
Events	399 (39.9)	260 (13.9)	610 (39.3)	1,269 (28.7)	
Cancer-specific death					<0.001
Censored	745 (74.4)	1,671 (89.5)	1,096 (70.7)	3,512 (79.5)	
Events	256 (25.6)	195 (10.5)	455 (29.3)	906 (20.5)	

 Table 1 (continued)

Data are presented as n (%). IAC, invasive adenocarcinoma; AJCC, American Joint Committee on Cancer.

Table 2 Demographics and clinicopathological characteristics in training cohort and validation cohort

Characteristic	Training cohort (n=3,092)	Validation cohort (n=1,326)	Total (N=4,418)	P value
Age, years				0.104
<60	513 (16.6)	212 (16.0)	725 (16.4)	
60–69	1,050 (34.0)	420 (31.7)	1,470 (33.3)	
70–79	1,124 (36.4)	486 (36.7)	1,610 (36.4)	
≥80	405 (13.1)	208 (15.7)	613 (13.9)	
Sex				1
Female	1,853 (59.9)	795 (60.0)	2,648 (59.9)	
Male	1,239 (40.1)	531 (40.0)	1,770 (40.1)	
Race				0.497
White	2,488 (80.5)	1,082 (81.6)	3,570 (80.8)	
Black	259 (8.4)	112 (8.4)	371 (8.4)	
Other	345 (11.2)	132 (10.0)	477 (10.8)	
Marital status				0.685
Married	1,842 (59.6)	773 (58.3)	2,615 (59.2)	
Divorced	393 (12.7)	166 (12.5)	559 (12.6)	
Single	419 (13.6)	181 (13.7)	600 (13.6)	
Widowed	438 (14.2)	206 (15.5)	644 (14.6)	
Derived AJCC T stage				0.807
T1	1,536 (49.7)	674 (50.8)	2,210 (50.0)	
T2	768 (24.8)	316 (23.8)	1,084 (24.4)	
Т3	246 (8.0)	99 (7.5)	345 (7.8)	
T4	542 (17.5)	237 (17.9)	779 (17.6)	

Table 2 (continued)

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Table 2	(continued)
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Characteristic	Training cohort (n=3,092)	Validation cohort (n=1,326)	Total (N=4,418)	P value
Derived AJCC N stage				0.137
N0	2,504 (81.0)	1,053 (79.4)	3,557 (80.5)	
N1	220 (7.1)	98 (7.4)	318 (7.2)	
N2	285 (9.2)	148 (11.2)	433 (9.8)	
N3	83 (2.7)	27 (2.0)	110 (2.5)	
Derived AJCC M stage				0.722
M0	2,700 (87.3)	1,152 (86.9)	3,852 (87.2)	
M1	392 (12.7)	174 (13.1)	566 (12.8)	
Derived AJCC stage				0.902
1	1,886 (61.0)	812 (61.2)	2,698 (61.1)	
II	334 (10.8)	134 (10.1)	468 (10.6)	
111	480 (15.5)	206 (15.5)	686 (15.5)	
IV	392 (12.7)	174 (13.1)	566 (12.8)	
Differentiation				0.678
High	692 (22.4)	309 (23.3)	1,001 (22.7)	
Median	1,303 (42.1)	563 (42.5)	1,866 (42.2)	
Low	1,097 (35.5)	454 (34.2)	1,551 (35.1)	
Surgery				0.253
Yes	2,435 (78.8)	1,023 (77.1)	3,458 (78.3)	
No	657 (21.2)	303 (22.9)	960 (21.7)	
Death				0.697
Censored	2,198 (71.1)	951 (71.7)	3,149 (71.3)	
Events	894 (28.9)	375 (28.3)	1,269 (28.7)	
Cancer-specific death				1
Censored	2,458 (79.5)	1,054 (79.5)	3,512 (79.5)	
Events	634 (20.5)	272 (20.5)	906 (20.5)	

Data are presented as n (%). AJCC, American Joint Committee on Cancer.

the training cohort and testing cohort showed no significant differences (P>0.05, chi-square test) in demographics or clinical characteristics (*Table 2*).

Exploration of prognostic risk factors associated with OS and CSS

By analyzing univariable and multivariable data, risk factors associated with the survival of IAC patients in the training

cohort were identified. On univariate analysis, older age, male sex, white, lower differentiated, divorced or widowed, higher T/N/M stage, higher TNM stage, and no surgery predicted poorer OS and CSS (*Table 3*). Moreover, results are also presented in the form of Kaplan-Meier curves (*Figures 2,3*). All variables analyzed were significant and were included in multivariate analysis. In the multivariate analysis of both OS and CSS, variables including age, sex, race, differentiation, marital stage, TNM stage and

Table 3 Univariate Cox analysis in OS and CSS

Variable		OS			CSS	
variable	HR	95% CI	P value	HR	95% CI	P value
Age, years						
<60			Reference			Reference
60–69	1.030	0.853-1.245	0.756	0.932	0.755–1.151	0.51
70–79	1.538	1.287–1.839	<0.001	1.218	0.995–1.489	0.04
≥80	2.490	2.056-3.016	<0.001	1.982	1.594–2.465	<0.001
Sex						
Female			Reference			Reference
Male	1.476	1.322-1.648	<0.001	1.541	1.353–1.755	<0.001
Race						
White			Reference			Reference
Black	0.741	0.593–0.9268	0.009	0.731	0.560-0.955	0.021
Other	0.680	0.559–0.827	<0.001	0.733	0.584–0.919	0.007
Differentiation						
High			Reference			Reference
Median	0.734	0.626-0.863	<0.001	0.791	0.654–0.958	0.016
Low	1.413	1.243–1.605	<0.001	1.571	1.346-1.834	< 0.001
Marital status						
Married			Reference			Reference
Divorced	1.154	0.973–1.368	0.1	1.12	0.917-1.369	0.267
Single	0.815	0.675-0.984	0.03	0.742	0.591–0.933	< 0.001
Widowed	1.433	1.242–1.653	<0.001	1.321	1.113–1.569	< 0.001
Derived AJCC T	stage					
T1			Reference			Reference
T2	1.382	1.181–1.618	<0.001	1.613	1.320-1.972	< 0.001
Т3	2.070	1.656-2.586	<0.001	2.753	2.114-3.585	< 0.001
T4	3.929	3.450-4.476	<0.001	5.961	5.078-6.999	< 0.001
Derived AJCC N	stage					
N0			Reference			Reference
N1	1.891	1.542-2.320	<0.001	2.278	1.810-2.867	<0.001
N2	3.801	3.277-4.407	<0.001	4.818	4.088–5.679	<0.001
N3	5.705	4.505-7.224	<0.001	7.107	5.487-9.205	<0.001
Derived AJCC M	stage					
M0			Reference			Reference
M1	5.955	5.280-6.717	<0.001	8.176	7.139–9.363	< 0.001

Table 3 (continued)

Veriable		OS			CSS			
Variable	HR	95% CI	P value	HR	95% CI	P value		
Derived AJCC stage								
I			Reference			Reference		
II	1.473	1.169–1.857	<0.001	1.634	1.209–2.209	0.001		
III	2.656	2.296-3.071	<0.001	3.894	3.256-4.656	<0.001		
IV	7.957	6.952-9.107	<0.001	12.850	10.911-15.121	<0.001		
Surgery								
Yes			Reference			Reference		
No	4.932	4.411–5.514	<0.001	6.078	5.323–6.941	<0.001		

Table 3 (continued)

OS, overall survival; CSS, cancer-specific survival; HR, hazard ratio; CI, confidence interval; AJCC, American Joint Committee on Cancer.

surgery were all statistically significant (*Table 4*) and were determined as independent risk factors for OS and CSS of IAC. However, the differences between blacks and whites in race group were not significant. Analysis of multivariate data found that the prognosis improved in patients of younger age, female, other-race (Asians, Pacific Islanders and Hispanics), highly differentiated stage, lower TNM stage, and surgery.

Subgroup analysis of the effects of differentiation

The effects of differentiation on CSS and OS regarding other prognostic factors were evaluated (Figures 4,5). Differences in differentiation were significant in the 60-69 years group (P<0.0001, Figure 4B; P<0.0001, Figure 5B) and the 70-79 group (P<0.0001, Figure 4C; P<0.0001, Figure 5C). We previously speculated the effect of differentiation on prognosis should be more and more significant with the increase of age, but there was no remarkable difference in the \geq 80 group (P=0.78, Figure 4D; P=0.98, Figure 5D), which may due to the limited cases in this group, and other higher-risk clinical conditions among the elderly. In addition, significant differences in female group (P<0.0001, Figure 4E; P<0.0001, Figure 5E), male group (P<0.0001, Figure 4F; P<0.0001, Figure 5F), white-race group (P<0.0001, Figure 4K; P<0.0001, Figure 5K), surgery group (P<0.0001, Figure 4R; P<0.0001, Figure 5R), and non-surgery group (P<0.0001, Figure 4S; P<0.0001, Figure 5S). To be noted, the results revealed that the contribution of differentiation to prognosis increased with higher TNM stage (Figures 4G-47,5G-57).

Furthermore, the OS and CSS curves of high, moderate, and low differentiation groups did not make a distinction in stage I (P=0.097, *Figure 4G*; P=0.92, *Figure 5G*). In stage II, the 10-year CSS of lowly differentiated patients was just over half that of highly differentiated patients (*Figure 5H*). We also acknowledged that the OS and CSS of low-differentiation IAC patients were lower than that of the other two groups in stage III and IV (*Figures 4I-47, 5I-57*).

Construction of the nomogram for OS and CSS

All significant risk factors of Cox regression in the training group were included in the differentiation-specific prognostic nomograms. Besides, tumor size was added in nomograms due to the importance of IAC T-stage and forest maps were produced as a result of the final multivariate analysis (Figure 6). Based on these independent prognostic factors of high, median, and low differentiated IAC, the differentiation-specific nomogram models for 1-, 3- and 5-year IAC patients' OS and CSS were constructed (Figure 7). The nomograms indicated that TNM stage was the strongest prognostic factor both in OS and CSS. Older age could indicate worse OS rather than CSS and we guess that old people are more likely to die from other diseases. The marital status played more important roles in highdifferentiation nomograms (Figure 7A,7D) compared with others (Figure 7B, 7C, 7E, 7F). Surgery had a great impact on each nomogram, while race played only a small role. The predicted probabilities of OS and CSS in 1, 3 and 5 years were obtained by consolidating the scores associated with



tumor-node-metastasis.



Table 4 Multivariate Cox analysis in OS and CSS

Voriable		OS			CSS		
vanable	HR	95% CI	P value	HR	95% CI	P value	
Age, years							
<60			Reference			Reference	
60–69	1.28	1.06–1.56	0.011	1.25	1.01–1.56	<0.001	
70–79	1.75	1.45–2.11	<0.001	1.49	1.21–1.84	<0.001	
≥80	2.23	1.81–2.75	<0.001	1.91	1.5–2.43	<0.001	
Sex							
Female			Reference			Reference	
Male	1.48	1.32–1.67	<0.001	1.51	1.31–1.73	<0.001	
Race							
White			Reference			Reference	
Black	0.933	0.743–1.17	0.549	0.905	0.689–1.19	0.471	
Other	0.728	0.597–0.887	<0.001	0.765	0.608-0.962	0.022	
Differentiation							
High			Reference			Reference	
Median	0.909	0.772-1.07	<0.001	0.97	0.799–1.18	0.76	
Low	1.53	1.34–1.75	<0.001	1.65	1.41–1.94	<0.001	
Marital status							
Married			Reference			Reference	
Divorced	1.3	1.09–1.55	0.003	1.28	1.04–1.57	0.017	
Single	1.02	0.845–1.24	0.803	0.971	0.768–1.23	<0.001	
Widowed	1.4	1.2–1.65	<0.001	1.39	1.15–1.69	<0.001	
Derived AJCC stage							
I			Reference			Reference	
Ш	1.44	1.14–1.82	<0.001	1.59	1.17–2.16	0.003	
III	2.37	2.04-2.75	<0.001	3.39	2.82-4.06	<0.001	
IV	4.78	4.07–5.61	<0.001	7.29	6.01-8.85	<0.001	
Surgery							
Yes			Reference			Reference	
No	2.54	2.22-2.92	<0.001	2.69	2.28-3.18	<0.001	

OS, overall survival; CSS, cancer-specific survival; HR, hazard ratio; CI, confidence interval; AJCC, American Joint Committee on Cancer.

each variable and converting them to the bottom scales.

Validation of the nomograms

As a way to demonstrate the advantages of the nomogram

models, ROC curves were plotted to identify accurate predictability for OS and CSS in the training and validation cohorts, respectively (*Figure 8*). AUC values of the ROC predicting the 1-, 3- and 5-year OS (*Figure 8A-8C*) and CSS (*Figure 8D-8F*) in training cohort were statistically



Figure 4 OS curves of IAC patients according to differentiation in different subgroups. (A-D): age. (A) <60 years; (B) 60–69 years; (C) 70–79 years; (D) \geq 80 years. (E,F): sex. (E) female; (F) male. (G-J): TNM stage. (G) stage I; (H) stage II; (I) stage III; (J) stage IV. (K-M): race. (K) white people; (L) black people; (M) other (Asian, Pacific Islander or American Indian). (N-Q): marital status. (N) married; (O) single; (P) divorced; (Q) widowed. (R,S): surgery. (R) surgery; (S) no-surgery. OS, overall survival; IAC, invasive adenocarcinoma; TNM, tumor-node-metastasis.

satisfactory, as well as in the validation cohort (*Figure 8G-8L*). AUC values of OS and CSS in the training cohort were 0.817 [95% confidence interval (CI): 0.753-0.881] and 0.835 (95% CI: 0.774-0.896), while in the validation cohort were 0.784 (95% CI: 0.695-0.874) and 0.813 (95% CI: 0.722-0.905). Moreover, to investigate the discrimination of models constructed, patients were separated into high-risk and low-risk groups by median of model scores. As was revealed, the high-risk groups always had lower survival than the low-risk groups (P<0.001, Log-rank test) in both training cohorts (*Figure 9A-9F*) and validation cohorts (*Figure 9G-9L*). On

each nomogram, a calibration curve was plotted to assess its accuracy. Calibration plot for predicting OS and CSS of 1-, 3- and 5-year demonstrated good predictability in *Figure 10*. Meanwhile, models for predicting 1-, 3- and 5-year survival of IAC were better superior to those in non-treated and all-treated groups in DCA curves (*Figure 11*).

Discussion

Nowadays, lung cancer is still at the top place for cancerrelated mortality, and LUAD is becoming more and more



Figure 5 CSS curves of IAC patients according to differentiation in different subgroups. (A-D): age. (A) <60 years; (B) 60–69 years; (C) 70–79 years; (D) \geq 80 years. (E,F): sex. (E) female; (F) male. (G-J): TNM stage. (G) stage I; (H) stage II; (I) stage III; (J) stage IV. (K-M): race. (K) white people; (L) black people; (M) other (Asian, Pacific Islander or American Indian). (N-Q): marital status. (N) married; (O) single; (P) divorced; (Q) widowed. (R,S): surgery. (R) performed; (S) none. CSS, cancer-specific survival; IAC, invasive adenocarcinoma; TNM, tumor-node-metastasis.

frequent (1,15). IAC plays an important role in LUAD but only a few researches have been conducted on IAC and its variants. In view of a few researches on the relationship between pathological grades and other factors of prognosis, this study focused on the connection and developed nomograms to estimate the outcomes of IAC, which could contribute to the development of a strategic treatment.

The study found that older patients, especially those over 80 years old, had distinctly lower OS and CSS. Simultaneously, age had a strong effect on nomograms (*Figure* 7), especially in median-differentiation nomogram models (*Figure* 7*B*,7*E*). Several studies have shown that advanced age has a detrimental effect on prognoses of IAC patients, which consisted with the findings in this study (2,3).

LUAD cases have increased dramatically over the past 20 years, especially among females (23). Therefore, the dataset of this study was composed mainly of female patients and the female gender could be a protective factor of IAC. Marital status has been shown to significantly influence the prognosis of LUAD patients since divorced patients are less likely



Figure 6 Final multivariate analysis for IAC. (A) Multivariate Cox analysis in OS. (B) Multivariate Cox analysis in CSS. HR, hazard ratio; CI, confidence interval; TNM, tumor-node-metastasis; IAC, invasive adenocarcinoma; OS, overall survival; CSS, cancer-specific survival.

to enjoy a relatively harmonious family environment (24). In this study, marital status for the first time was found to be an independent risk factor for CSS and OS of IAC, which was added to the nomogram. Studies have shown that white patients had the worst survival rate among IAC patients of different races (8,9), which consisted with our results. Moreover, as with most solid tumors, surgery remains the most effective treatment for IAC. According to our data, OS and CSS were significantly prolonged in surgical patients compared to non-surgical patients. Most patients undergoing surgery were in stage I (88.5%), II (90.9%), and III (75.9%). We speculated that the lower-staged patients were in better physical condition and at younger age, which may contribute to a better prognosis. Only 22.0% of stage IV patients opted for surgery and the prognosis (cancer-related deaths and noncancer-related deaths) was poor in this group.

Histological patterns have long been recognized as an independent prognostic predictor of LUAD, closely related to biological characteristics and surgical outcomes (25-28). Our study divided the histological patterns into three types

of differentiation and confirmed that the prognostic role of histological patterns still existed in IAC.

It is generally believed that lowly differentiated LUAD has a high recurrence rate and poor prognosis. Liu et al. (29) reported that patients with lowly differentiated LUAD had a higher risk of recurrence than those with highly differentiated LUAD throughout the follow-up. Notably, the recurrence hazard curve in low-differentiated patients showed a typical "double-peaked" pattern, which means the recurrence of patients was concentrated at 20-22 months and 5-6 years after surgery, while the curve of high-differentiated patients is relatively smooth after surgery. Some types of low-differentiated IAC, such as micropapillary and solid, are significantly associated with lymph node metastasis (30). Besides, pathological differentiation is an independent predictor for local recurrence, distant metastasis, chest recurrence and brain recurrence (26,29,31). IASLC pathology committee has specified a histology-based grading system for invasive pulmonary adenocarcinoma, which plays an important



Figure 7 Nomograms for predicting the 1-, 3-, and 5-year OS and CSS of IAC with different differentiation. (A) OS nomogram for IAC of high-differentiation; (B) OS nomogram for IAC of median-differentiation; (C) OS nomogram for IAC of low-differentiation; (D) CSS nomogram for IAC of high-differentiation; (E) CSS nomogram for IAC of median-differentiation; (F) CSS nomogram for IAC of low-differentiation. OS, overall survival; IAC, invasive adenocarcinoma; TNM, tumor-node-metastasis; CSS, cancer-specific survival.

role in prognosis stratification (17,32-34). To be noted, IAC is histologically and molecularly heterogeneous even in patients with the same TNM stage, which affects their prognosis and treatment strategies (35). Therefore, to comprehensively predict the prognosis of ICA, other factors, such as TNM stage, age, and surgery should also be

considered along with pathological differentiation. Patients with high- and moderate-differentiation had a better OS and CSS than low-differentiation patients at stage III and IV, while the differences was not significant in stage I patients. Some studies have explored the relationship between pathological components and prognosis after



Figure 8 ROC curves and AUC values for nomogram models. (A-C) ROC curves and AUC values for 1-, 3- and 5-year OS in different differentiation training cohorts. (D-F) ROC curves and AUC values for 1-, 3- and 5-year CSS in different differentiation training cohorts. (G-I) ROC curves and AUC values for 1-, 3- and 5-year OS in different differentiation cohorts. (J-L) ROC curves and AUC values for 1-, 3- and 5-year OS in different differentiation cohorts. (J-L) ROC curves and AUC values for 1-, 3- and 5-year OS in different differentiation cohorts. (J-L) ROC curves and AUC values for 1-, 3- and 5-year CSS in different differentiation cohorts. AUC, area under curve; ROC, receiver operating characteristic; OS, overall survival; CSS, cancer-specific survival.



Figure 9 Kaplan-Meier curves of OS and CSS based on risk-level. (A-C) OS by risk-level with differentiation in training cohort; (D-F) CSS by risk-level with different differentiation in training cohort; (G-I) OS by risk-level with different differentiation in validation cohort; (J-L) CSS by risk-level with different differentiation in validation cohort. OS, overall survival; CSS, cancer-specific survival.



Figure 10 Calibration plots of OS and CSS associated nomograms in both training and validation cohorts. The dashed line represents perfect agreement between the nomogram-predicted probability (x-axis) and the actual probability, calculated from a Kaplan-Meier analysis (y-axis). (A-C) Calibration plots of 1-, 3- and 5-year OS in training cohort; (D-F) calibration plots of 1-, 3- and 5-year CSS in training cohort; (G-I) calibration plots of 1-, 3- and 5-year OS in validation cohort; (J-L) calibration plots of 1-, 3- and 5-year CSS in validation cohort; OS, overall survival; CSS, cancer-specific survival.



Figure 11 DCA for 1-, 3- and 5-year. (A) DCA curve of high-differentiation for 1-, 3-, and 5-year OS. (B) DCA curve of mediandifferentiation for 1-, 3-, and 5-year OS. (C) DCA curve of low-differentiation for 1-, 3-, and 5-year OS. (D) DCA curve of highdifferentiation for 1-, 3-, and 5-year CSS. (E) DCA curve of median-differentiation for 1-, 3-, and 5-year CSS. (F) DCA curve of lowdifferentiation for 1-, 3-, and 5-year CSS. DCA, decision curve analysis; OS, overall survival; CSS, cancer-specific survival.

different types of surgery, revealing that the micropapillary/ solid subtype indicates worse prognosis overall (36-38). Differentiation is correlated to almost all factors, except young, lower TNM-stage, non-white and widowed patients. Thus, for personalized medicine paradigm, clinicians can classify IAC patients by level of differentiation to predict prognosis. Histological subtypes have also been proved to be an independent factor of adjuvant chemotherapy, while micropapillary/solid subtype was considered as a negative predictor (28,39). Some doctors suggest patients with low differentiated IAC should receive postoperative adjuvant therapy (40). Due to the unavailability of chemoradiotherapy data, they are not included in our study.

Based on former studies, the combination of pathological differentiation and other factors has great influence on the prognosis of IAC, caused by complicated mechanisms. Many studies have focused on this aspect with no certain conclusion reached. However, there are some clues supporting pathogenesis, progression and prognosis of IAC are influenced by genomic alteration, especially in different pathological subtypes. Compared with high-differentiated IAC (LEP) and moderate-differentiated IAC (ACI/PAP),

previous studies revealed that tumor mutational burden (TMB), fraction of genome altered (FGA), copy number amplifications, whole-genome doubling, rate of transversion and number of altered oncogenic pathways were higher in low-differentiated IAC (MIP/SOL and some variants) (41,42). The median of TMB and FGA increased as the subtypes became more aggressive. Besides, various altered genes have been found in distinct differentiated groups of lung cancer. Three genes (EGFR, RBM10, and TERT) are significantly altered in LEP, while 4 genes (TP53, SETD2, MGA, and SMARCA4) are significantly altered in MIP/ SOL. Studies show that SOLs have the greatest amount of transversion/transition, while lepidic adenocarcinomas have the lowest amount. Mutations in PI3K pathways are closely related to recurrence and metastasis in ACI, PAP, MIP, and SOL subtypes (IAC with moderate and low differentiation). In pathologic subtypes that show higher aggression, EGFR mutations are less frequent, while the frequency of cell cycle pathway changes including TP53 mutations goes up (43). Some studies reported that KRAS and EGFR mutations are highly correlated with the invasiveness and inertia of tumor tissue, respectively (44). Adenocarcinomas and mucinous

carcinomas were significantly characterized by KRAS mutations, the presence of tumor-infiltrating leukocytes and heavy smoking history. On the contrary, there was an association between EGFR mutations combined with nonsmoking history and high/median-differentiation IAC, especially lepidic and papillary.

In recent years, immunotherapy has given new hope to lung cancer patients and have achieved good clinical results in some cases (45,46). Different pathological subtypes also shared disparate different immune profiles. Solid predominant IAC often showed the overexpression of PD-L1 and a high percentage of tumor-infiltrating lymphocytes, which provided a potential target for treatment-adjuvant PD-1 blockade immunotherapy. Dong et al. (42) claimed that the pathological subtype of IAC could serve as a potential predictor of adjuvant immunotherapy. Among IAC patients, the co-expression of PD-L1/CD47 has a prognostic effect, associated with an increase in CD8⁺ T-cells (47). Therefore, it is believed that PD-L1/CD47 co-expression could be used to predict the efficacy of dual-targeting immunotherapy. Based on the deepening understanding of different immune characteristics, pathology-specific immunotherapy may achieve good clinical results in the future.

Nomograms are graphical displays of mathematical models, which integrate biological and clinical variables to forecast the probability of certain clinical events. Although nomograms are a reliable tool for prognosis prediction, to some extent, the model is still not perfect in this study. First, this is a retrospective observational study, therefore inherent selection bias is inevitable, which suggests that further prospective comparative studies are needed. Second, the nomogram has not included some critical factors, such as smoking history, lymphovascular invasion, predominant growth patterns, and mixed pathological subtypes, due to the unavailability of this information in the SEER database. Rather than a distinct subtype of pathology, IACs are generally mixture of heterogeneous subtypes and the patterns are usually continuous (e.g., lepidic to papillary or acinar). The SEER database provides only one subtype, which could result in a decrease in the universality of the nomogram. In addition, many other factors that may affect prognosis, such as mutations of the KRAS and EGFR genes and expression of PD-L1. Third, the potential interaction terms are not considered in the nomogram to improve the conciseness and interpretability. Taking all potential interaction terms in model construction may lead to better predictive performance but this will make the model more

complex and difficult to use in clinical practice. Despite validating the data, clinical trials could have provided more convincing evidence for the conclusion. In addition to the data of SEER database, the accuracy of models could be further judged by follow-up of patients with clear pathology.

Notably, this appears to be the first research to separate the prognostic factors of IAC based on differentiation level, followed by nomogram model construction and evaluation. Nomogram accuracy can be evaluated by using the AUCs of ROC curves and calibration curves. In our nomogram models, the majority of AUCs predicting the 1-, 3- and 5-year OS and CSS are higher than 0.75, suggesting high model accuracy. The calibration curve and DCA curves are also satisfactory, which means our nomogram models are highly reliable.

Conclusions

This study explored the impact of disparate pathological differentiation on the prognosis of IAC and developed differentiation-specific nomogram models to predict the OS and CSS in 1-, 3- and 5-year. Using these nomogram models, clinicians can accurately predict the prognosis and select appropriate treatment options for IAC patients. In future studies, we will further explore the pathogenesis and genomic changes of IAC with different pathological differentiation.

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Footnote

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Ethical Statement: The authors are accountable for all aspects of the work in ensuring that questions related to the accuracy or integrity of any part of the work are appropriately investigated and resolved. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). Since the study relied on the SEER database, which was freely available to the public (http:// seer.cancer.gov/seerstat/), the ethical approval was waived.

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