

# Tumour-pleura relationship on computed tomography (CT) provides effective risk stratification for peripheral pulmonary nodules with Lung Imaging Reporting and Data System (Lung-RADS) score of 4X

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**Background:** Pulmonary nodules with Lung Imaging Reporting and Data System (Lung-RADS) 4X are of greater clinical significance, and accurate differentiation of pathological types and visceral pleural invasion (VPI) of Lung-RADS 4X peripheral pulmonary nodules before treatment can aid in stratification. This study set out to investigate whether the tumour-pleura relationship on computed tomography (CT) can provide effective risk stratification for peripheral pulmonary nodules with Lung-RADS 4X.

**Methods:** This was a single institution, retrospective study of 482 consecutive patients with Lung-RADS score 4X, who were pathologically diagnosed with tuberculous granuloma and adenocarcinoma from January 2019 to December 2023. We assessed clinical factors (baseline characteristics and tumour markers) and CT findings. Univariate and multivariate logistic regression analyses were used to determine the classification of pulmonary nodules and predictors of VPI.

**Results:** Multivariate analysis revealed that gender [odds ratio (OR) =0.392; P<0.001], carcinoembryonic antigen (CEA) level (OR =8.331; P<0.001), type of nodules (OR =13.551 and 7.478; P<0.001 and P=0.016) and maximum base width of soft tissue component on the pleura side (OR =0.857; P=0.005) were significant independent factors for distinguishing tuberculous granuloma from adenocarcinoma. And the type of linear connection between lesion and pleura (OR =3.936; P<0.001), and the maximum base width of soft tissue components on the pleura side (OR =1.359; P=0.001) were correlated independently with VPI. The area under the curve (AUC) for predicting pulmonary nodules classification was 82.60% [95% confidence interval (CI): 78.85–86.35%), and the AUC for predicting VPI was 76.10% (95% CI: 69.83–82.38%).

**Conclusions:** The tumour-pleura relationship will be helpful in further risk stratification for peripheral pulmonary nodules with a score of Lung-RADS 4X.

<sup>\*</sup>These authors contributed equally to this work as co-first authors.

**Keywords:** Lung neoplasms; pleura; biomarkers; X-ray computed tomography (X-ray CT)

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## Introduction

Low-dose computed tomography (LDCT) is the primary screening method for early detection of lung cancer and can reduce mortality in certain high-risk populations (1,2). To standardize the image reporting of pulmonary nodules and management of pulmonary nodule patients, the American College of Radiology developed the Lung Imaging Reporting and Data System (Lung-RADS) based on the characteristics of pulmonary nodules (such as size, attenuation, and growth pattern) to stratify the risk of lesions (3). Category 4X is a special category of lesions that exhibit additional features and imaging findings and represents the highest risk of malignant tumours, with a malignancy rate greater than 15% (4,5). In addition, most of the lung cancer detected by screening belong to pulmonary nodules with a Lung-RADS score of 4X, which makes these nodules highly clinically significant (6).

Among Lung-RADS 4X pulmonary nodules, the most common pathological type of benign pulmonary nodule is tuberculous granuloma, which is challenging to distinguish due to its similar imaging features to lung cancer (7). Accurate identification of the two can guide clinicians to make reasonable and effective treatment and prevent patients from receiving overtreatment or losing the optimal treatment time due to delayed treatment (8). Lung adenocarcinoma is the most common pathological type of peripheral lung cancer, and visceral pleural invasion (VPI) is an important prognostic factor and staging index (9). For patients with early-stage peripheral lung adenocarcinoma, the five-year survival rate is 86% for patients without VPI, compared with 57-70% for patients with VPI (10). In addition, in the presence of VPI, extensive surgery and adjuvant chemotherapy can significantly improve patient survival outcomes (11-13). Therefore, accurate differentiation of pathological types and VPI of Lung-RADS 4X peripheral pulmonary nodules before treatment can aid in stratification, and maximize the guarantee that patients receive reasonable and accurate treatment.

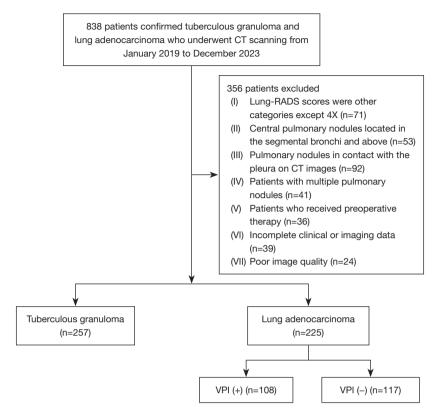
Several previous studies have reported that the

relationship between peripheral lung cancer and the pleura plays an important role in further risk stratification of lesions (9,14). However, whether the relationship between peripheral pulmonary nodules of Lung-RADS 4X and the pleura can provide an important basis for classifying pulmonary nodules and VPI has not been explored by scholars. Therefore, the aim of this study was to confirm whether the relationship between lesion and pleura could provide valid risk stratification for Lung-RADS 4X peripheral pulmonary nodules. We present this article in accordance with the STROBE reporting checklist (available at https://qims.amegroups.com/article/view/10.21037/qims-24-530/rc).

## **Methods**

## Patients

This retrospective study was approved by the ethics committee of Lanzhou University Second Hospital (No. 2023A-745), and the requirement for informed consent was waived. The study was conducted in accordance with the Declaration of Helsinki (as revised in 2013). A total of 838 patients with pathologically confirmed diagnosis of tuberculous granuloma and lung adenocarcinoma who underwent computed tomography (CT) scanning from January 2019 to December 2023 were retrospectively enrolled in this study. The exclusion criteria were as follows: (I) Lung-RADS scores for pulmonary nodules of other categories except 4X; (II) central pulmonary nodules located in the segmental bronchi and above (main bronchi, lobar bronchi); (III) pulmonary nodules in contact with the pleura on CT images; (IV) patients with multiple pulmonary nodules; (V) patients who received preoperative anti-inflammatory, antituberculosis or antitumor therapy; (VI) incomplete clinical or imaging data; (VII) poor image quality not meeting the research requirements. Finally, a total of 482 patients were included in our study. The inclusion and exclusion criteria are shown in Figure 1. The Lung-RADS v2022 is used for the management of indeterminate pulmonary nodules. For pulmonary nodules with Lung-RADS score of 4X, the following is



**Figure 1** Flow chart of patients who enrolled in this study. CT, computed tomography; Lung-RADS, Lung Imaging Reporting and Data System; VPI, visceral pleural.

recommended: (I) diagnostic chest CT with or without contrast; (II) positron emission tomography (PET)/CT may be considered if there is a ≥8 mm (≥268 mm³) solid nodule or solid component; (III) tissue sampling; (IV) and/or referral for further clinical evaluation. Clinical data including gender, age and lung cancer tumour markers [carcinoembryonic antigen (CEA), neuron-specific enolase (NSE), cytokeratin fragment 19 (CYFRA21-1), pro-gastrin-releasing peptide (ProGR), and squamous cell carcinoma antigen (SCCA)] were collected.

# CT protocol

All patients underwent preoperative chest CT scanning using multi-slice spiral CT scanners (Discovery CT 750 HD and Revolution CT, GE Healthcare, USA; Somatom Force, Siemens Healthcare, Germany; iCT 256, Philips Healthcare, Amsterdam, USA). The patient was trained to breath before the scan and was instructed to hold his or her breath as long as possible after inhalation during the scan. The patient was placed in the supine

position on the CT scanning bed, head first, with arms straight and raised above the head. CT scanning range was as follows: thoracic entrance to the lower margin of diaphragm. The tube voltage was 120 kVp, the tube current range was 150–200 mm, the scanning layer thickness and scanning spacing were both 5 mm, and the reconstruction layer thickness and layer spacing were both 1.25 mm.

# Image analysis

Two chest radiologists with 6 and 10 years of experience independently collected clinical data and evaluated the CT images without knowing the histopathological results. The following CT data were collected: (I) type of nodule (solid, subsolid and ground-glass nodule); (II) maximum diameter of the nodule; (III) maximum diameter of solid component; (IV) type of linear connection between lesion and pleura (single line or multiple lines); (V) length of the line connecting the lesion to the pleura (shortest line length measured on axial, sagittal or coronal images); (VI) type of adjacent pleura (interlobular pleura, peripheral

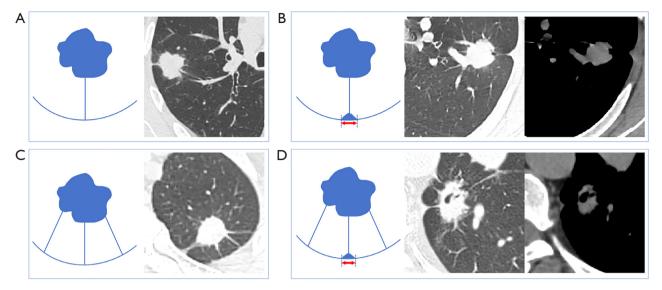


Figure 2 Schematic and example of tumour-pleura relationships. (A) Single linear pleural tags on lung window; (B) single linear or striated pleural tags with soft tissue component at the pleural side were visible; (C) multiple linear pleural tags on lung window; (D) multiple linear or striated pleural tags with soft tissue component at the pleural side were visible. The red arrows in (B) and (D) are the maximum base width of soft tissue.

pleura or both); (VII) presence of soft tissue components on the pleura side, and measurement of the maximum base width of soft tissue components on the pleura side if present (*Figure 2*). All CT images were observed on the lung and mediastinal windows. In case of disagreements during the analysis, the authors would discuss and come to a consensus.

# Statistical analysis

SPSS 26.0 (IBM Corp, USA), R software (version 4.2.6) and MedCalc software (version 4.2.6) were used for statistical analysis. Kappa and the intraclass correlation coefficient (ICC) were used to analyze the consistency between classifications and continuous variables obtained by two radiologists. The Shapiro-Wilk test was used to analyze the normality of continuous variables. Two-sample t-test and Mann-Whitney U test were used to compare the differences between the two groups of continuous variables. The Chisquare or Fisher's exact probability test was used to compare the difference between the two groups of categorical variables. Multivariate logistic regression model was used to screen for statistically significant variables, a predictive model was developed and the model was visualized as a nomogram. ROC curves were used to evaluate the predictive efficiency of each independent predictor and the combined model.

## **Results**

# Patient characteristics

Table 1 summarizes the detailed information of the 482 patients with Lung-RADS 4X, 257 patients with tuberculous granuloma and 225 patients with lung adenocarcinoma were included in our study, where in lung adenocarcinoma included 108 patients with presence of VPI and 117 patients without VPI. This study included 282 males and 200 females with a mean age of 57.90±10.51 years. Gender and CEA level were significantly different between tuberculous granuloma and adenocarcinoma patients (P<0.001), while age and other tumour markers (NSE, CYFRA21-1, ProGR, and SCCA) were not significantly different between the two groups (P>0.05). In terms of VPI, only CEA was significantly different between the two groups (P=0.021).

# Univariate analysis of conventional CT signs

The interobserver agreement between the two radiologists in analyzing the CT signs was good ( $\kappa$ =0.92–0.99; ICC =0.86–0.94). In the differentiation of tuberculous granuloma and adenocarcinoma with Lung-RADS 4X, there were significant differences in type of nodules, type of linear connection between lesion and pleura, presence of soft

Table 1 Baseline demographic and clinical characteristics of the study population

Maxiala I a a	Tuberculous granuloma	Lung adenocarcinoma	Lung adenocar	D4	P2	
Variables	(n=257)	(n=225)	VPI (+) (n=108)			- P1
Gender					<0.001	0.410
Male	178 (69.3)	104 (46.2)	53 (49.1)	51 (43.6)		
Female	79 (30.7)	121 (53.8)	55 (50.9)	66 (56.4)		
Age (years)	53.00 (45.00, 60.50)	58.00 (52.50, 65.00)	57.06±8.47	57.06±8.47 59.07±9.61		0.098
Tumor markers						
CEA					<0.001	0.021
Positive	22 (8.6)	61 (27.1)	38 (35.2)	25 (21.4)		
Negative	235 (91.4)	164 (72.9)	70 (64.8)	92 (78.6)		
NSE					0.273	0.402
Positive	23 (8.9)	27 (12.0)	15 (13.9)	12 (10.3)		
Negative	234 (91.1)	198 (88.0)	93 (86.1)	105 (89.7)		
ProGRP					0.949	0.142
Positive	52 (20.2)	45 (20.0)	26 (24.1)	19 (16.2)		
Negative	205 (79.8)	180 (80.0)	82 (75.9)	98 (83.8)		
SCCA					1.000	1.000
Positive	5 (1.9)	4 (1.8)	2 (1.9)	2 (1.7)		
Negative	252 (98.1)	221 (98.2)	106 (98.1)	115 (98.3)		
CYFRA21-1					0.649	0.233
Positive	34 (13.2)	33 (14.7)	19 (17.6)	14 (12.0)		
Negative	223 (86.8)	192 (85.3)	89 (82.4)	103 (88.0)		
N stage					_	0.12
N0	-	-	94 (87.0)	109 (93.2)		
N1	-	-	14 (13.0)	8 (6.8)		
M stage					-	0.27
M0	-	-	99 (91.7)	112 (95.7)		
M1	-	-	9 (8.3)	5 (4.3)		

Normally distributed variables were expressed as mean ± standard deviations and non-normally distributed variables as medians (Q1, Q3), categorical variables were expressed as count (percentage). CEA, carcinoembryonic antigen; NSE, neuron-specific enolase; ProGR, progastrin-releasing peptide; SCCA, squamous cell carcinoma antigen; CYFRA21-1, cytokeratin fragment 19; VPI, visceral pleural invasion; P1, tuberculous granuloma vs. lung adenocarcinoma; P2, visceral pleural invasion positive vs. visceral pleural invasion negative.

tissue components on the pleura side and the maximum base width of soft tissue components on the pleura side (P values were <0.001, 0.001, <0.001, and <0.001, respectively). There were no significant differences in the maximum diameter of nodules, the maximum diameter of solid components, length of the line connecting the lesion to the pleura, the

type of adjacent pleura and CT signs of lung nodule itself (calcification, cavity, air bronchogram, lobulation sign, spiculation, bubblelike lucency, necrosis and CT value) between them (P>0.05) (*Table 2*).

In predicting VPI, there were statistically significant differences in the type of nodule, the maximum diameter

Table 2 Comparison of imaging features between two groups

Variables	Tuberculous granuloma	Lung adenocarcinoma	Lung adenocar	rcinoma (n=225)	P1	P2	
variables	(n=257)	(n=225)	VPI (+) (n=108)	VPI (-) (n=117)	FI		
Type of nodules					<0.001	0.005	
Solid	243 (94.6)	129 (57.3)	71 (65.7)	58 (49.6)			
Subsolid	12 (4.7)	85 (37.8)	36 (33.3)	49 (41.9)			
Ground-glass	2 (0.8)	11 (4.9)	1 (0.9)	10 (8.5)			
Maximum diameter of the nodule (mm)	19.41 (16.54, 23.23)	19.41 (17.37, 23.01)	23.89±6.59	21.88±8.46	0.906	0.049	
Calcification					0.11	0.99	
Yes	44 (17.1)	27 (12.0)	13 (12.0)	14 (12.0)			
No	213 (82.9)	198 (88.0)	95 (88.0)	103 (88.0)			
Cavity					0.84	0.28	
Yes	36 (14.0)	33 (14.7)	13 (12.0)	20 (17.1)			
No	221 (86.0)	192 (85.3)	95 (88.0)	97 (82.9)			
Air bronchogram					0.51	0.43	
Yes	52 (20.2)	51 (22.7)	22 (20.4)	29 (24.8)			
No	205 (79.8)	174 (77.3)	86 (79.6)	88 (75.2)			
Lobulation sign					0.26	0.09	
Yes	117 (45.5)	114 (50.7)	61 (56.5)	53 (45.3)			
No	140 (54.5)	111 (49.3)	47 (43.5)	64 (54.7)			
Spiculation					0.07	0.33	
Yes	103 (40.1)	109 (48.4)	56 (51.9)	53 (45.3)			
No	154 (59.9)	101 (51.6)	52 (48.1)	64 (54.7)			
Bubblelike lucency					0.66	0.36	
Yes	101 (39.3)	84 (37.3)	37 (34.3)	47 (40.2)			
No	156 (60.7)	141 (62.7)	71 (65.7)	70 (59.8)			
Necrosis					0.49	0.37	
Yes	29 (11.3)	23 (10.2)	9 (8.3)	14 (12.0)			
No	228 (88.7)	202 (89.8)	99 (91.7)	103 (88.0)			
CT value (HU)	37 (32, 43)	36 (33, 42)	37.72±7.13	36.06±5.97	0.13	0.07	
Maximum diameter of solid component (mm)	18.96 (16.05, 23.01)	18.78 (12.00, 23.51)	20.86 (15.86, 26.3)	16.27 (10.28, 20.98)	0.093	<0.001	
Type of linear connection					0.001	<0.001	
Single line	37 (14.4)	59 (26.2)	16 (14.8)	53 (45.3)			
Multiple line	220 (85.6)	166 (73.8)	92 (85.2)	64 (54.7)			
Length of the line (mm)	7.63 (4.95, 10.47)	6.87 (4.88, 10.21)	6.87 (4.88, 10.21)	7.63 (4.95, 10.47)	0.082	0.599	

Table 2 (continued)

Table 2 (continued)

Variables	Tuberculous granuloma	Lung adenocarcinoma	Lung adenocar	Lung adenocarcinoma (n=225)		
variables	(n=257)	(n=257)		· P1	P2	
Type of adjacent pleura					0.166	0.036
Interlobular pleura	12 (4.7)	15 (6.7)	8 (7.4)	7 (6.0)		
Peripheral pleura	225 (87.5)	183 (81.3)	81 (75.0)	102 (87.2)		
Both	3oth 20 (7.8)		27 (12.0) 19 (17.6)			
Soft tissue on the pleura	side				< 0.001	0.001
Positive	237 (92.2)	181 (80.4)	97 (89.8)	84 (71.8)		
Negative	20 (7.8)	44 (19.6)	11 (10.2)	33 (28.2)		
Maximum base width of soft tissue (mm)	4.83 (3.22, 7.10)	3.39 (2.23, 5.45)	4.31 (3.02, 6.25)	2.62 (0.00, 4.45)	<0.001	<0.001

Normally distributed variables were expressed as mean ± standard deviations and non-normally distributed variables as medians (Q1, Q3), categorical variables were expressed as count (percentage). VPI, visceral pleural invasion; P1, tuberculous granuloma vs. lung adenocarcinoma; P2, visceral pleural invasion positive vs. visceral pleural invasion negative; CT, computed tomography; HU, Hounsfield unit; Maximum base width of soft tissue, maximum base width of soft tissue components on the pleura side; Type of linear connection, type of linear connection between lesion and pleura; Length of the line, length of the line connecting the lesion to the pleura; Soft tissue on the pleura side, presence of soft tissue components on the pleura side.

of nodule, the maximum diameter of solid component, type of linear connection between lesion and pleura, type of adjacent pleura, presence of soft tissue components on the pleura side and the maximum base width of soft tissue components on the pleura side between them (P values were 0.005, 0.049, <0.001, <0.001, 0.036, 0.001 and <0.001, respectively), but there was no significant difference in length of the line connecting the lesion to the pleura and CT signs of lung nodule itself (calcification, cavity, air bronchogram, lobulation sign, spiculation, bubblelike lucency, necrosis and CT value) between the two (P>0.05) (*Table 2*).

In addition, type of nodule, type of linear connection between lesion and pleura, presence of soft tissue components on the pleura side and the maximum base width of soft tissue components on the pleura side were statistically different both in the classification of pulmonary nodules and VPI (P<0.05).

# Multivariate logistic regression analysis

Multivariate logistic regression analysis revealed that gender [odds ratio (OR) =0.392; P<0.001], CEA level (OR =8.331; P<0.001), type of nodule (OR =13.551 and 7.478; P<0.001 and P=0.016) and the maximum base width of soft tissue components on the pleura side (OR =0.857; P=0.005) were independent risk factors for distinguishing

lung adenocarcinoma from tuberculous granuloma. The variance inflation factor for the above six potential predictors ranged from 1.005–1.089, indicating the absence of multicollinearity. A nomogram was constructed using the six predictors above to provide a visual measure of outcomes (*Table 3, Figure 3A*).

Multivariate logistic regression analysis was used to identify two independent predictors with variance inflation factors of 1.142 and 2.053. Multivariate analysis revealed that the type of linear connection between lesion and pleura (OR =3.936; P<0.001) and the maximum base width of soft tissue components on the pleura side (OR =1.359; P=0.001) were independently associated with VPI. A nomogram was constructed to visualize the results (*Table 4*, *Figure 3B*).

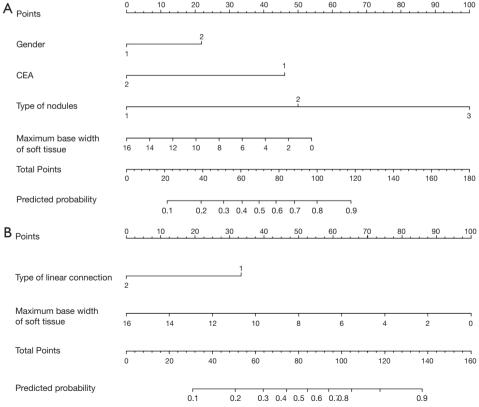
# Diagnostic performance evaluation

ROC curve analysis showed that the area under the curve (AUC) of gender and CEA in differentiating adenocarcinoma from tuberculous granuloma was 61.52% [95% confidence interval (CI): 56.47–66.56%] and 60.30% (95% CI: 55.20–65.41%), respectively. The AUC of type of nodules and the maximum base width of soft tissue components on the pleura side were 68.58% (95% CI: 63.70–73.45%) and 65.14% (95% CI: 60.26–70.03%), respectively. When the maximum base width of soft tissue

Table 3 Multivariate logistic regression analysis for predictive factors for distinguishing tuberculous granuloma from adenocarcinoma

M. California	Mu		
Variables	OR	95% CI	– P
Gender			
Male (vs. female)	0.392	0.247-0.620	<0.001
CEA			
Positive (vs. negative)	8.331	4.486–15.472	<0.001
Type of nodules			
Subsolid (vs. solid)	13.551	6.871–26.727	<0.001
Ground-glass (vs. solid)	7.478	1.450–38.555	0.016
Type of linear connection			
Multiple (vs. single)	0.627	0.358-1.099	0.103
Soft tissue on the pleura side			
Positive (vs. negative)	1.08	0.444-2.630	0.865
Maximum base width of soft tissue (mm)	0.857	0.769-0.954	0.005

OR, odds ratio; CI, confidence interval; CEA, carcinoembryonic antigen; Type of linear connection, type of linear connection between lesion and pleura; Soft tissue on the pleura side, presence of soft tissue components on the pleura side; Maximum base width of soft tissue, maximum base width of soft tissue components on the pleura side.



**Figure 3** Nomogram of two models. (A) Nomogram for distinguishing tuberculous granuloma from adenocarcinoma developed based on gender, CEA, type of nodules, maximum base width of soft tissue; (B) nomogram for predicting visceral pleural invasion developed based on type of linear connection and maximum base width of soft tissue. CEA, carcinoembryonic antigen.

Table 4 Multivariate logistic regression analysis for predictive factors for predicting visceral pleural invasion

Variables	Mu	ltivariable	D
Variables	OR	95% CI	— Р
CEA			
Positive (vs. negative)	1.362	0.691–2.686	0.374
Type of nodules			
Subsolid (vs. solid)	0.839	0.407-1.728	0.633
Ground-glass (vs. solid)	0.305	0.028-3.342	0.331
Maximum diameter of the nodule	0.977	0.928-1.028	0.374
Maximum diameter of solid component (mm)	1.010	0.977-1.044	0.570
Type of linear connection			
Multiple (vs. single)	3.936	1.926-8.044	<0.001
Type of adjacent pleura			
Peripheral (vs. interlobular)	0.425	0.121-1.490	0.181
Both (vs. interlobular)	0.912	0.198-4.208	0.907
Soft tissue on the pleura side			
Positive (vs. negative)	1.827	0.580-5.755	0.303
Maximum base width of soft tissue (mm)	1.359	1.135–1.628	0.001

OR, odds ratio; CI, confidence interval; CEA, carcinoembryonic antigen; Type of linear connection, type of linear connection between lesion and pleura; Soft tissue on the pleura side, presence of soft tissue components on the pleura side; Maximum base width of soft tissue, maximum base width of soft tissue components on the pleura side.

components on the pleura side was 3.16, the sensitivity and specificity were 45.78% (95% CI: 39.1–52.5%) and 78.21% (95% CI: 72.7–83.1%), respectively. A logistic regression combined model was established for the above statistically significant variables, and the AUC was 82.60% (95% CI: 78.85–86.35%%) (*Table 5, Figure 4*).

The AUC of the type of linear connection between lesion and pleura and the maximum base width of soft tissue components on the pleura side for predicting VPI were 65.24% (95% CI: 58.07–72.41%) and 71.31% (95% CI: 64.53–78.09%), respectively. When the maximum base width of soft tissue components on the pleura side was 3.51, the sensitivity and specificity were 66.67% (95% CI: 56.9–75.4%) and 71.79% (95% CI: 62.7–79.7%), respectively. A Logistic regression combined model was established with the AUC of 76.10% (95% CI: 69.83–82.38%) (*Table 6*, *Figure 4*).

## **Discussion**

In this study, we first assessed the classification of Lung-

RADS 4X peripheral pulmonary nodules using clinical factors and the tumour-pleura relationship, and then further evaluated VPI in patients with Lung-RADS 4X peripheral lung adenocarcinoma using the above clinical and imaging factors. Our study found that gender, CEA, type of nodule, and the maximum base width of soft tissue components on the pleura side were independent risk factors for distinguishing lung adenocarcinoma from tuberculous granuloma. In addition, the type of linear connection between lesion and pleura and the maximum base width of soft tissue components on the pleura side were independent predictors of VPI in patients with peripheral lung adenocarcinoma. ROC curve analysis revealed that the combined model based on the above independent predictors demonstrated good predictive performance in predicting the classification and VPI of Lung-RADS 4X peripheral pulmonary nodules (AUC were 82.60% and 76.10%, respectively).

CEA is a reproducible and widely used tumour marker. Previous studies have shown that CEA enhances intercellular adhesion and promotes tumour formation

Table 5 Predictive performance for distinguishing tuberculous granuloma from adenocarcinoma of the single and combined predictive factors

Variables	AUC (95% CI), %	P value	ΥI	Cut-off	Sensitivity (95% CI), %	Specificity (95% CI), %	PLR (95% CI)	NLR (95% CI)
Gender	61.52 (56.47–66.56)	<0.001	0.23	-	53.78 (47.0–60.4)	69.26 (63.2–74.8)	1.75 (1.4–2.2)	0.67 (0.6–0.8)
CEA	60.30 (55.20–65.41)	<0.001	0.21	-	28.00 (22.2–34.4)	96.21 (88.7–95.5)	3.79 (2.3–6.1)	0.78 (0.7–0.8)
Type of nodules	68.58 (63.70–73.45)	<0.001	0.37	-	42.67 (36.1–49.4)	94.55 (91.0–97.0)	7.83 (4.6–13.3)	0.61 (0.5–0.7)
Maximum base width of soft tissue (mm)	65.14 (60.26–70.03)	<0.001	0.24	3.16	45.78 (39.1–52.5)	78.21 (72.7–83.1)	2.10 (1.6–2.8)	0.69 (0.6–0.8)
Combined	82.60 (78.85–86.35)	<0.001	0.54	0.46	70.67 (64.2–76.5)	83.27 (78.1–87.6)	4.22 (3.2–5.6)	0.35 (0.3-0.4)

AUC, area under the curve; CI, confidence interval; YI, Youden Index; PLR, positive likelihood ratio; NLR, negative likelihood ratio; CEA, carcinoembryonic antigen; Maximum base width of soft tissue, maximum base width of soft tissue components on the pleura side.

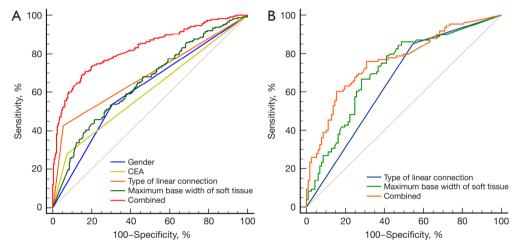


Figure 4 ROC curve analysis of two models. (A) Predictive performance for distinguishing tuberculous granuloma from adenocarcinoma of the combined predictive factors, gender, CEA, type of nodules, maximum base width of soft tissue based on AUC analysis; (B) predictive performance for predicting visceral pleural invasion of the combined predictive factors, type of linear connection and maximum base width of soft tissue based on AUC analysis. CEA, carcinoembryonic antigen; ROC, receiver operating characteristic; AUC, the area under the curve.

Table 6 Predictive performance for predicting visceral pleural invasion of the single and combined predictive factors.

Variables	AUC (95% CI), %	P value	ΥI	Cut-off	Sensitivity (95% CI), %	Specificity (95% CI), %	PLR (95% CI)	NLR (95% CI)
Type of linear connection	65.24 (58.07–72.41)	<0.001	0.30	-	85.19 (77.1–91.3)	45.30 (36.1–54.8)	1.56 (1.3–1.9)	0.33 (0.2–0.5)
Maximum base width of soft tissue (mm)	71.31 (64.53–78.09)	<0.001	0.38	3.51	66.67 (56.9–75.4)	71.79 (62.7–79.7)	2.36 (1.7–3.2)	0.46 (0.3–0.6)
Combined	76.10 (69.83–82.38)	<0.001	0.45	0.50	75.93 (66.7–83.6)	69.23 (60.0–77.4)	2.47 (1.8–3.3)	0.35 (0.2–0.5)

AUC, area under the curve; CI, confidence interval; YI, Youden Index; PLR, positive likelihood ratio; NLR, negative likelihood ratio; Maximum base width of soft tissue, maximum base width of soft tissue components on the pleura side.

in a variety of tumour cells, and is widely used in the detection of a variety of tumours, especially in the diagnosis and prognosis of lung cancer (15-17). The results of this study showed that CEA was significantly different in the classification of pulmonary nodules and VPI, and was an independent predictor of tuberculous granuloma and lung adenocarcinoma. Several researchers have shown that CEA levels are greater in lung cancer patients than those in patients with tuberculous granuloma, especially those with lung adenocarcinoma (8,18). Moreover, Zhu *et al.* (19) and Fujibayashi *et al.* (20) reported that CEA was a good marker for predicting VPI in non-small cell lung cancer. This is consistent with our findings.

The results of our study suggest that type of nodule and the maximum base width of soft tissue components on the pleura side are independent risk factors for distinguishing Lung-RADS 4X lung adenocarcinoma from tuberculous granuloma. In terms of type of nodules, although both can present as solid, subsolid or ground-glass nodules, tuberculous granuloma is more often solid, whereas lung adenocarcinoma is more often subsolid and ground-glass nodules. Henschke et al. (21) reported that patients with subsolid nodules (63%) had a higher risk of malignancy, followed by those with ground-glass nodules (18%) and solid nodules (7%). Kim et al. (22) also showed that approximately 75% of ground-glass nodules were lung adenocarcinoma. The presence of a soft tissue component at the pleural side correlates with the type of peripheral lung nodule, and lung adenocarcinomas are due to the growth of tumour cells towards the pleura, thus filling the cavity between the depressed visceral pleura and the wall pleura. Tuberculous granuloma is more likely to involve the pleura, causing exudation and fibrin deposition (23). Therefore, we hypothesized that tuberculous granuloma is more likely to cause pleural adhesion thickening than lung adenocarcinoma, resulting in more extensive formation of soft tissue components on the pleural side. However, in order to confirm our results, further studies in a wider population are needed.

VPI is one of the key determinants of pathological staging of lung cancer and is associated with occult lymph node metastasis and local recurrence (24,25). The results of this study showed that the type of linear connection between lesion and pleura and the maximum base width of soft tissue components on the pleura side were independent predictors of VPI in Lung-RADS 4X peripheral lung adenocarcinoma. The linear connection between lesions and pleura is thought to be thickened interlobular interval

extending from the surface of the nodule to the surface of the pleura, and may be related to local oedema, the spread of tumour cells in and out of the lymphatic vessels, inflammatory cells or fibrosis (9). Therefore, the likelihood of VPI is greater for multilinear connections. This is consistent with the findings of Yang et al. (26) and Hsu et al. (9). We hypothesized that the maximum base width of soft tissue components on the pleura side was related to the amount of tumour cells entering the interpleural space between the visceral and mural layers, and that the more aggressive the tumour was, the more extensive the soft tissue formation on the pleural side was, the greater the soft tissue base width will be.

In this study, although some variables were significantly different between the two groups in the univariate analysis, they were not significantly different in multivariate logistic regression. Such as those predicting VPI including type of nodules, maximum diameter of the nodule, maximum diameter of solid component, type of adjacent pleura and presence of soft tissue components on the pleura side. This is because multivariate logistic regression analysis balances the effects of confounding factors and allows screening for objective factors that predict the type of pulmonary nodule and VPI.

This study has several limitations: First, as a single-center retrospective study, we only selected patients who underwent surgery for the evaluation of VPI, which is inevitable for potential selection bias. Second, due to the low percentage of other pathological types such as fungal infection, inflammatory lesions, squamous cell lung cancer, and small cell lung cancer among Lung-RADS 4X peripheral pulmonary nodules, they were not included in this study, and in the future, we need to include multicenter and more pathological types for research to obtain more extensive and accurate conclusions.

### **Conclusions**

In conclusion, for Lung-RADS 4X peripheral pulmonary nodules, the type of nodules and the maximum base width of soft tissue components on the pleura side were found to be independent risk factors for distinguishing lung adenocarcinoma from tuberculous granuloma, and the type of linear connection between lesion and pleura and the maximum base width of soft tissue components on the pleura side were independently associated with VPI. These easy-to-use quantitative variables will help to provide effective risk stratification of Lung-RADS 4X peripheral

pulmonary nodules.

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#### **Footnote**

Reporting Checklist: The authors have completed the STROBE reporting checklist. Available at https://qims.amegroups.com/article/view/10.21037/qims-24-530/rc

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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). The study was approved by the ethics committee of Lanzhou University Second Hospital (No. 2023A-745), and the requirement for informed consent was waived for this retrospective study.

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