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*CORRESPONDENCE Zhuoyue Tang Zhuoyue_tang@cqmu.edu.cn

[†]These authors have contributed equally to this work

RECEIVED 18 December 2023 ACCEPTED 02 May 2024 PUBLISHED 28 May 2024

CITATION

Ren X, Zhang J, Song Z, Li Q, Zhang D, Li X, Yu J, Li Z, Wen Y, Zeng D, Zhang X and Tang Z (2024) Detection of malignant lesions in cytologically indeterminate thyroid nodules using a dual-layer spectral detector CTclinical nomogram. *Front. Oncol.* 14:1357419. doi: 10.3389/fonc.2024.1357419

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Detection of malignant lesions in cytologically indeterminate thyroid nodules using a dual-layer spectral detector CT-clinical nomogram

Xiaofang Ren^{1,2†}, Jiayan Zhang^{1,2†}, Zuhua Song², Qian Li², Dan Zhang², Xiaojiao Li², Jiayi Yu², Zongwen Li², Youjia Wen², Dan Zeng², Xiaodi Zhang³ and Zhuoyue Tang^{1,2*}

¹Department of Radiology, Affiliated Hospital of Southwest Medical University, Luzhou, China, ²Department of Radiology, Chongqing General Hospital, Chongqing, China, ³Department of Clinical and Technical Support, Philips Healthcare, Chengdu, China

Purpose: To evaluate the capability of dual-layer detector spectral CT (DLCT) quantitative parameters in conjunction with clinical variables to detect malignant lesions in cytologically indeterminate thyroid nodules (TNs).

Materials and methods: Data from 107 patients with cytologically indeterminate TNs who underwent DLCT scans were retrospectively reviewed and randomly divided into training and validation sets (7:3 ratio). DLCT quantitative parameters (iodine concentration (IC), NIC_P (IC nodule/IC thyroid parenchyma), NIC_A (IC nodule/IC ipsilateral carotid artery), attenuation on the slope of spectral HU curve and effective atomic number), along with clinical variables, were compared between benign and malignant cohorts through univariate analysis. Multivariable logistic regression analysis was employed to identify independent predictors which were used to construct the clinical model, DLCT model, and its performance was assessed using receiver operating characteristic curve, calibration curve, and decision curve analysis. The nomogram was subsequently tested in the validation set.

Results: Independent predictors associated with malignant TNs with indeterminate cytology included NIC_P in the arterial phase, Hashimoto's Thyroiditis (HT), and BRAF V600E (all p < 0.05). The DLCT-clinical nomogram, incorporating the aforementioned variables, exhibited superior performance than the clinical model or DLCT model in both training set (AUC: 0.875 vs 0.792 vs 0.824) and validation set (AUC: 0.874 vs 0.792 vs 0.779). The DLCT-clinical nomogram demonstrated satisfactory calibration and clinical utility in both training set and validation set.

Conclusion: The DLCT-clinical nomogram emerges as an effective tool to detect malignant lesions in cytologically indeterminate TNs.

KEYWORDS

thyroid nodule, cytology, multidetector computed tomography, nomograms, diagnosis

Introduction

Thyroid nodules (TNs) are highly prevalent, being detected in up to 65% of the general population (1). The nature of TN is the primary concern for both patients and clinicians, as it defines subsequent clinical management. Ultrasound-guided fine-needle aspiration biopsy (US-FNAB) has been regarded as the preoperative standard method for assessing the state of TN (2). However, approximately 10-25% of patients who underwent US-FNAB had cytologically indeterminate TNs, including Bethesda categories III, IV, and V with risks of malignancy of 10-30%, 25-40%, and 50-75%, respectively (2-4). This subset of patients would be recommended for diagnostic lobectomy due to the inability to rule out cancer, with an inevitable consequence of unnecessary and potentially harmful surgery in benign nodules or completion thyroidectomy following the initial diagnostic lobectomy in malignant nodules (2). Studies have shown that approximately 60% of these patients underwent unnecessary surgery (5). Therefore, a reliable preoperative method that could better detect malignant lesions in cytologically indeterminate TN is sorely needed for appropriate clinical management to reduce the burden on patients and health systems.

Some potential imaging tools have been investigated in the differential diagnosis of cytologically indeterminate TN, such as US and positron emission tomography (PET)/CT (6–8). US is the primary imaging method for risk stratification of TN with the advantage of being accessible, inexpensive, and nonradiative (9). However, US has some limitations in assessing the nature of TN due to similarities in ultrasound features between benign and malignant lesions (10), diminished objectivity for operator dependence (11), and various diagnostic standards between different US risk stratification systems (12). (18F)PET-CT showed a high predictive value for malignancy (13), but it is expensive and not widely available. Thus, an objective, accessible, and reliable tool for the proper management of cytologically indeterminate TN is still needed.

Dual-layer detector spectral CT (DLCT) is an advanced functional imaging technique that could reflect valuable hemodynamic and tissue characterization for lesion detection by offering multiparametric data not obtained by conventional CT imaging (14, 15). This method has the advantage of increasing the sensitivity and qualitative accuracy of lesion detection and minimizing metallic artifacts (16). Frequently used DLCT-derived quantitative parameters, including iodine concentration (IC), the slope of the spectral HU curve (λ_{HU}), and effective atomic number (Z_{eff}), have demonstrated diagnostic potential value in oncologic applications (17, 18). Additionally, previous studies have supported the use of DLCT for differentiating between benign TN and thyroid cancer and evaluating lymph node metastases in thyroid cancer (19, 20). However, the value of quantitative parameters from DLCT in

detecting malignant lesions in cytologically indeterminate TN remains elusive.

In this study, we hypothesize that DLCT-derived parameters may be a reliable method to detect malignant TNs with indeterminate cytology. To test our hypothesis, we sought to develop a nomogram combining DLCT parameters and clinical variables to detect malignant TNs in cytologically indeterminate TNs to reduce the need for diagnostic surgery.

Materials and methods

Patients selection

This retrospective study received approval from the institutional review board of Chongqing General Hospital, and the written informed consent requirement was waived.

Between August 2021 and January 2023, we reviewed the records of 182 patients who underwent DLCT scans and thyroid surgery for indeterminate thyroid nodules. Exclusion criteria were: (1) borderline tumor confirmed by surgery; (2) thyroid nodules could not be identified on DLCT images; (3) the nodules were extensive calcification, cystic change or necrosis; (4) incomplete clinical information or images; (5) patients had a history of malignancy. Out of the 182 patients initially considered, 75 were excluded. Ultimately, 107 TNs with indeterminate cytology from 107 patients were included and randomly divided into training set and validation set at a ratio of 7:3. All of them were finally pathologically confirmed after surgery. The training cohort was used for model building, while the validation cohort was used for internal validation of the model. The flowchart of patient selection was shown in Figure 1.

Image acquisition

All patients in our study underwent preoperative contrastenhanced neck scans through 64-slice DLCT devices (IQon Spectral CT, Philips Healthcare). The DLCT scanning protocol was as follows: tube voltage, 120 kVp; tube current, modulated with automated exposure control (DoseRight system, Philips Healthcare); detector collimation, 64×0.625 mm; field of view, 350 mm; matrix, 512×512 ; layer thickness, 3 mm; reconstruction thickness, 0.67 mm. Patients were intravenously injected with nonionic contrast media (Iopamidol 350 mg/ml, Bracco) at a dose of 1.5 ml/kg and an injection rate of 3.5 ml/s, followed by 30 ml of saline flashing at the same rate. The arterial phase (AP) scan was performed with a delay of 6 seconds and a venous phase (VP) scan was performed with a delay of 50 seconds after the CT value of the descending aorta lumen at the tracheal bifurcation level reached 150 Hounsfield units, respectively.

Indeterminate TNs identification

The location and size of indeterminate TNs were used as matching information to correlate nodules seen on US-guided FNAB with CT

Abbreviations: AUC, Area under receiver operating characteristic curve; AP, Arterial phase; BMI, Body mass index; CI, Confidence interval; DLCT, Dual-layer detector spectral computed tomography; FNAB, Guided fine-needle aspiration biopsy; HT, Hashimoto's thyroiditis; IC, Iodine concentration; ICC, Interclass correlation coefficient; IQR, Interquartile range; OR, Odds ratio; ROI, Region of interest; TN, Thyroid nodule; US, Ultrasound; VP, Venous phase; Zeff, Effective atomic number; λ_{HU} , The slope of the spectral HU curve.



images. The specific matching method was as follows: (1) The location (left lobe, right lobe, isthmus and superior, middle, inferior) of indeterminate TNs were acquired according to the cytologic pathology report; (2) The indeterminate TNs on CT images were identified by a radiologist who referred to the cytologic pathology report; (3) If several nodules lay close to one another, they were discriminated by size acquired from the US-FNAB report.

DLCT quantitative analysis

DLCT images were transferred from the picture archiving and communication system to a spectral CT postprocessing workstation (IntelliSpace Portal Version 10.1, Philips Healthcare) and were reconstructed into iodine image, monochromatic image, and effective atomic number image for subsequent quantitative analysis.

The approach for quantitative analysis is displayed as follows: First, radiologists enlarged images for accurately measuring lesions, ensuring a clear image. Secondly, circular regions of interest (ROIs) were manually drawn as large as possible on the solid area of the TN on axial iodine images avoiding calcification, cystic or necrotic regions, and additional ROIs of thyroid tissues and that of the ipsilateral carotid artery on the cross-sectional map were selected for normalization. Meanwhile, ROIs were automatically propagated to monochromatic images and effective atomic numbers. All quantitative parameters were measured twice to calculate the average. The measurements were taken by two independent observers trained in head and neck imaging who were only aware of the location of the lesion. Quantitative parameter-derived DLCT included IC of TN, thyroid parenchyma, and ipsilateral carotid artery, CT values on 40 keV and 100 keV monoenergetic images, and Z_{eff} . All quantitative parameters were measured in the AP and VP. To minimize variations caused by the patient's circulation or hormone status, the IC values of the ipsilateral adjacent artery and thyroid parenchyma were measured to derive the following ratios respectively: NIC_A = IC nodule/IC ipsilateral carotid artery, NIC_P = IC nodule/IC thyroid parenchyma. CT values on the monoenergetic images were measured to derive the following ratio: $\lambda_{HU} = (HU \ 40 \ keV - HU100 \ keV)/(100 - 40).$

Clinical data

Clinical variables were included in this study, such as age, sex, BRAF V600E, HT, and body mass index (BMI). BRAF V600E mutation was assessed by FNAB or the medical record. HT was diagnosed according to US and anti-thyroid antibodies (TPOAb > 30 IU/L or TgAb > 95 IU/L) (21). BMI was calculated by the following ratio: BMI = weight (kg)/height (m)².

Development and validation of nomogram

Statistically significant clinical variables and DLCT quantitative parameters between the benign nodule cohort and malignant nodule cohort were first identified using univariate analysis. Next, the significant variables from the univariate analysis were entered into multivariable logistic stepwise regression analyses to identify independent predictors for malignant TN with indeterminate cytology and to estimate their respective regression coefficients. Afterward, a DLCT-clinical model was developed, incorporating all selected independent predictors weighted by their respective regression coefficients. Similarly, a clinical model was constructed using independent clinical variables and a DLCT model was constructed using independent DLCT quantitative parameters.

The diagnostic ability of the three aforementioned models was evaluated through receiver operating characteristic curve and was tested in the validation set. Sensitivity and specificity were calculated. Decision curve analysis was adopted to assess the clinical utility of the three different models in both the training set and validation set. Subsequently, a nomogram was constructed from the model with optimal diagnostic performance. The calibration curve was applied to evaluate the fit goodness of the nomogram.

Statistical analysis

All statistical analyses were performed using R software (http:// www.R-project.org) and SPSS software (version 25.0, SPSS, IBM). For all DLCT parameters, interobserver reproducibility was assessed with intraclass correlation coefficients (ICC). The Shapiro-Wilk test was used to determine the distribution of continuous data. Normally distributed data were expressed as Mean ± Standard Deviation and assessed by a two-sample t-test, while non-normally distributed data were presented as Median (interquartile range, IQR) and analyzed by the Mann-Whitney U test. Categorical variables were presented as raw numbers, and analyzed by a chi-squared test. Multivariable logistic stepwise regression analyses were performed in the training set to identify independent predictors applied to prediction models and to estimate their odds ratio (OR) and a 95% confidence interval (CI). The stopping rule for forward stepwise selection was the likelihood ratio test with Akaike's information criterion. A twosided *p*-value < 0.05 indicated statistical significance.

Results

Patient characteristics

A total of 107 histologically confirmed TNs from 107 patients, comprising 44 benign and 63 malignant nodules, qualified for the final analysis of the study. According to the pathological results, the benign nodules included 32 nodular goiters, 3 follicular adenomas, 3 adenomatous nodular goiters, and 6 inflammatory nodules. The malignant nodules included 63 papillary carcinomas.

Univariate analysis and multivariable analysis

Characteristics of patients in the training and validation sets are detailed in Table 1. No significant differences in demographic and clinical characteristics were observed between the two sets (all p > 0.05).

Interclass correlation analysis for DLCT parameters showed good concordance between inter- and intraobserver agreements (ICC > 0.75). In the training set, DLCT parameters and clinical variables between the benign and malignant nodule cohorts are summarized in Table 2. Univariate analysis revealed significant differences in DLCT parameters (including IC_{nodule}, NIC_P, NIC_A, $\lambda_{\rm HU}$ and $Z_{\rm eff}$ in the AP and VP), BRAF V600E and HT between benign and malignant nodule cohorts (all p < 0.05). However, no significant differences were found in age (p = 0.171) between the benign cohort (mean age ± standard, 46.48 ± 11.71 years) and malignant cohort (mean age \pm standard deviation, 42.83 \pm 11.13 years) or in BMI (p = 0.806) between the benign cohort (median with the interquartile range, 21.97 with 21.03 - 24.05 kg/m²) and malignant cohort (median with the interquartile range, 22.32 with 20.83 - 23.90 kg/m²). Multivariable logistic stepwise regression analysis in the training set identified AP-NIC_p (OR = 0.004; 95% CI = 0.000 - 0.106; p < 0.001), HT (OR = 3.802; 95% CI = 1.037 -13.933; *p* = 0.044), and BRAF V600E (OR = 6.436; 95% CI = 1.618 -25.605; p = 0.008) as variables independently associated with malignancy (Table 3).

Predictive model analysis

Based on the selected predictors from multivariable logistic stepwise regression analysis, three predictive models were established: one clinical model (HT + BRAF V600E), one DLCT model (AP-NIC_p), and one DLCT-clinical combined model (HT + BRAF V600E + AP-NIC_p). The classification performance of the three models was shown in Table 4 and Figure 2. The AUC values were 0.792 (95% CI, 0.689 - 0.896) in the clinical model, 0.824 (95% CI, 0.729 - 0.919) in the DLCT model, and 0.875 (95% CI, 0.795 - 0.955) in the combined model for the training set. In the validation set, the AUC values were 0.792 (95% CI, 0.634 - 0.950) in the clinical model, 0.779 (95% CI, 0.608 - 0.950) in the DLCT model, and 0.874 (95% CI, 0.752 - 0.997) in the combined model.

Development and evaluation of the predictive nomogram

The DLCT-clinical nomogram was established according to the DLCT-clinical combined model (Figure 3). The formula was as follows: DLCT-clinical nomogram = $2.809 - 5.54 \times \text{APNIC}_{\text{p}} + 1.862 \times \text{BRAF V600E} + 1.335 \times \text{HT}$. Calibration curve of the nomogram demonstrated good agreement between predicted results and actual observations in both training and validation sets (Figures 4A, B). The decision curve analysis revealed that the nomogram provided more clinical benefit than the all-or-none intervention strategy and achieved optimal clinical utility among the three different models in both training and validation sets (Figures 4C, D). The DLCT and pathological images from the two examples were shown in Figures 5 and 6.

TABLE 1 Characteristics of patients in the training and validation sets.

| Characteristics | Training set (n = 75) | Validation set (n = 32) | P value |
|---------------------------------------|--------------------------|-------------------------------|------------|
| Age* | 44.44 ± 11.46 | 46.63 ± 11.89 | 0.374 |
| Sex | | | 0.256 |
| Male | 8 | 6 | |
| Female | 67 | 26 | |
| BMI (kg/m ²) [#] | 22.27 (20.96, 23.88) | 22.31 (20.52, 24.72) | 0.951 |
| BRAF V600E | | | 0.093 |
| mutant | 29 | 18 | |
| wild | 46 | 14 | |
| HT | | | 0.533 |
| yes | 33 | 12 | |
| no | 42 | 20 | |

Unless otherwise indicated, data are the number of patients. ^{*}data are mean \pm standard deviation. [#]data are median with the interquartile range in parentheses. BMI, body mass index; HT, Hashimoto's thyroiditis.

TABLE 2 Univariate analysis in the benign and malignant nodule cohort .

Discussion

In this study, a DLCT-clinical nomogram comprising AP-NIC_P, HT, and BRAF V600E demonstrated optimal diagnostic performance for detection of malignant lesions in cytologically indeterminate TN, outperforming both the DLCT model and clinical model in training and validation sets. These findings suggested that the DLCT-clinical nomogram could be a valuable tool for detection of malignant lesions in TNs with indeterminate cytology. Additionally, the relatively objective nature of the DLCTclinical nomogram made it user-friendly, eliminating the need for experience in identifying typical radiological features, which is especially beneficial for less experienced clinicians.

The study identified that adipose tissue could release cytokines and enhance oxidative stress, thereby promoting the progression of transformed cancer cells (22). However, we found that BMI, extensively used for assessing obesity, with an inability to detect malignant lesions in TNs with indeterminate cytology, which is consistent with previous studies (23–25). In our study, the clinical variables BRAF V600E and HT emerged as important independent predictors of malignant lesions in cytologically indeterminate TN, aligning with findings from prior studies (26–29). BRAF V600E, a

| Variables | Benign nodule cohort (n = 33) | Malignant nodule cohort (n = 42) | F value/Z value/X ² value | p value |
|--------------------------------|----------------------------------|-------------------------------------|---|---------|
| Age* | 46.48 ± 11.71 | 42.83 ± 11.13 | 0.012 | 0.171 |
| BMI (kg/m ²) | 21.97 (21.03, 24.05) | 22.32 (20.83, 23.90) | 22.32 (20.83, 23.90) -0.246 | |
| BRAF V600E [#] | | 13.740 | | <0.001 |
| mutant | 5 | 24 | | |
| wild | 28 | 18 | | |
| HT [#] | | | 4.487 | 0.034 |
| yes | 10 | 23 | | |
| no | 23 | 19 | | |
| AP | | | | |
| IC _{nodule} (mg/ml) | 3.84 (3.39, 4.35) | 2.33 (1.74, 2.92) | -4.910 | <0.001 |
| NIC _P | 0.84 (0.67, 0.93) | 0.53 (0.40, 0.71) | -4.792 | <0.001 |
| NICA | 0.40 (0.32, 0.46) | 0.24 (0.18, 0.32) | -5.113 | <0.001 |
| λ_{HU} | 4.66 (3.78, 5.38) | 2.88 (2.15, 3.64) | -4.531 | <0.001 |
| Z _{eff} * | 9.11 ± 0.41 | 8.59 ± 0.46 | 1.029 | <0.001 |
| VP | | | | |
| IC _{nodule} * (mg/ml) | 3.15 ± 0.88 | 2.70 ± 0.73 | 1.005 | 0.019 |
| NIC _P | 0.86 (0.74, 0.94) | 0.74 (0.65, 0.87) | -2.636 | 0.008 |
| NIC _A * | 0.81 ± 0.20 | 0.67 ± 0.22 | 0.106 | 0.008 |
| λ_{HU}^{*} | 3.88 ± 1.19 | 3.32 ± 0.90 | 1.399 | 0.026 |
| Z _{eff} * | 8.85 ± 0.36 | 8.67 ± 0.32 | 0.449 | 0.021 |

Unless otherwise indicated, data are median with the interquartile range in parentheses. *data are mean ± standard deviation. #data are the number of patients.

BMI, body mass index; HT, Hashimoto's thyroiditis; AP, arterial phase; VP, venous phase; IC_{nodule}, iodine concentration of thyroid nodule; NIC_P, IC_{nodule}/IC_{thyroid parenchyma}; NIC_A, IC_{nodule}/IC_{carotid artery}; λ_{HU} , slope of the spectral Hounsfield unit curve; Z_{eff}, effective atomic number.

TABLE 3 Stepwise multivariate logistic regression analysis of predictors associated with malignant thyroid nodule with indeterminate cytology in the training set.

| Variables | β value | OR | 95%CI | p value |
|---------------------|---------------|-------|--------------|---------|
| AP-NIC _P | -5.540 | 0.004 | 0.001-0.106 | 0.001 |
| НТ | 1.335 | 3.802 | 1.037-13.933 | 0.044 |
| BRAF V600E | 1.862 | 6.436 | 1.618-25.605 | 0.008 |

B. regression coefficient: OR, odds ratios: CL confidence interval: AP, arterial phase: NICp, ICnodule/ICthyroid parenchyma: HT, Hashimoto's thyroiditis.

TABLE 4 Results of three models' predictive ability for detecting malignant thyroid nodules with indeterminate cytology.

| Groups | Models | AUC | 95CI% | Sensitivity | Specificity |
|------------|---------------------|-------|-------------|-------------|-------------|
| Training | Clinical model | 0.792 | 0.689-0.896 | 0.881 | 0.576 |
| | DLCT model | 0.824 | 0.729-0.919 | 0.786 | 0.667 |
| | DLCT-clinical model | 0.875 | 0.795-0.955 | 0.833 | 0.727 |
| Validation | Clinical model | 0.792 | 0.634-0.950 | 0.857 | 0.545 |
| | DLCT model | 0.779 | 0.608-0.950 | 0.762 | 0.636 |
| | DLCT-clinical model | 0.874 | 0.752-0.997 | 0.857 | 0.727 |

AUC, area under the curve; CI, confidence interval; DLCT, dual-layer spectral detector computed tomography.

member of the Raf family of serine/threonine protein kinases, plays a pivotal role in cell proliferation, differentiation, and apoptosis (30). We propose that cells with BRAF V600E mutation may fail to undergo normal apoptosis, further triggering tumor occurrence (31). HT, an autoimmune thyroid disease, characterized by increased thyroid volume, lymphocyte infiltration of parenchyma, and the presence of antibodies specific to thyroid antigens (32), was associated with an increased probability of thyroid carcinoma in our study. This association may be attributed to lymphocytes stimulating cancer cell proliferation by secreting chemokines and other molecules (33). Moreover, prolonged elevated levels of TSH in HT patients may fuel follicular epithelial proliferation contributing to the development of PTC (32). The clinical model combining BRAF V600E with HT exhibited good sensitivity and relatively poor specificity (0.576 in the

training set and 0.545 in the validation set), suggesting that the clinical model had limitations in reducing inappropriate therapy for patients with cytologically indeterminate TNs.

The study further explored the diagnostic values of DLCT parameters and revealed that IC, λ_{HU} , NIC_A, NIC_P, and Z_{eff} in the AP and VP were significantly lower in the malignant cohort than in the benign cohort. Multivariable analysis identified AP-NIC_P as another independent predictor for malignant TN with indeterminate cytology. Tumor cells can disrupt thyroid follicular cells, which have an iodine uptake function, leading to a reduction in IC of the malignant lesion (34, 35). NICp, calculated based on IC, can provide quantitative information about iodine uptake and tissue perfusion level, superior to IC in overcoming the bias from patient hormone levels (16). Our results indicated that the DLCT model



FIGURE 2

Comparison of the different models for distinguishing benign from malignant thyroid nodules with indeterminate cytology in the training set (A) and validation set (B). DLCT, dual-layer spectral detector computed tomography



(AP-NICp) was useful for detection of TNs with indeterminate cytology, achieving AUCs of 0.824 and 0.779 in the training and validation sets, respectively. However, accurately identifying TN status remained challenging, with poor specificity (0.667 in the training set and 0.636 in the validation set).

Based on these above results, we established a DLCT-clinical model incorporating the clinical variables (HT and BRAF V600E) with AP-NICp to detect malignant lesions in cytologically indeterminate TN. The DLCT-clinical model demonstrated the best diagnostic utility with an AUC of 0.875 in the training set and an AUC of 0.874 in the validation set. Meanwhile, the combined model exhibited good sensitivity (0.857 in the validation set) and specificity (0.727 in the validation set). Moreover, the DLCT-clinical model was visually presented by a nomogram, serving as a simple tool for clinician use. Typical radiological features were excluded from our study due to the subjectivity of operator dependence. Our proposed nomogram exhibited good agreement with pathological results, as

shown in the calibration curves, and demonstrated good clinical utility for patients with cytologically indeterminate TN, as seen in decision curve. We compared the diagnostic performance of DLCT with different examinations (Supplementary Material). It showed that the DLCT-clinical nomogram was not inferior to some different examinations. Therefore, the DLCT-clinical nomogram may serve as a simple, objective, and reliable tool to guide clinical personalized assessment.

Studies reported that artificial intelligence algorithms allow for precise quantification of features such as nuclear area and elongation factor or crowding of nuclei, which appear to be differently distributed between benign and malignant nodules, revealing the potential to efficiently evaluate FNA cytology cases (36). The whole slide imaging, a digital pathology modality, facilitates the use of artificial intelligence in the field of pathology (37). In the future study, we will develop an artificial intelligence model with multidimensional features including radiological



FIGURE 4

Calibration curves of the DLCT-clinical nomogram in the training set (A) and validation set (B). Decision curve analysis of the three models in the training set (C) and validation set (D).



FIGURE 5

The DLCT quantitative parameters and Haematoxylin-eosin stain in a 37-year-old woman with thyroid papillary carcinoma who tested positive for HT and negative for BRAF. (**A**, **B**) CT value of arterial phase 40 keV and 100 keV monochromatic images is 173.5 HU, and 37.9 HU, respectively. (**C**) Arterial phase NIC_A and NIC_P are 0.18, and 0.48, respectively. (**D**) Arterial phase Z_{eff} is 8.29. (**E**) US images. (**F**, **G**) CT value of venous phase 40 keV and 100 keV monochromatic image is 199.2 HU, and 46.3 HU, respectively. (**H**) Venous phase NIC_A and NIC_P are 0.50, and 0.80, respectively. (**I**) Venous phase SEC (**F**, **G**) CT value of venous phase 40 keV and 100 keV monochromatic image is 199.2 HU, and 46.3 HU, respectively. (**H**) Venous phase NIC_A and NIC_P are 0.50, and 0.80, respectively. (**I**) Venous phase Zeff is 8.39. (**J**) Photomicrograph confirmed the pathological finding of the nodule as papillary thyroid carcinoma. (Hematoxylineosin stain; original magnification, 40). (**K**) Arterial phase λ_{HU} of the energy curve is 2.26 HU/keV. (**L**) Venous phase λ_{HU} of the energy curve is 2.55 HU/keV. HT, Hashimoto's thyroiditis; AP, arterial phase; VP, venous phase; HU, CT value; λ_{HU} , the slope of spectral HU curve; IC_{nodule}, iodine concentration of thyroid nodule; NIC_P, IC_{nodule}/IC_{hyroid} parenchyma; NIC_A, IC_{nodule}/IC_{carotid} artery; λ_{HU} , slope of the spectral Hounsfield unit curve; Z_{eff}, effective atomic number; US, Ultrasound; H&E, Haematoxylin-eosin.



FIGURE 6

The DLCT quantitative parameters and Haematoxylin-eosin stain in a 47-year-old woman with nodular goiter who tested negative for HT and positive for BRAF. (A, B) CT value of arterial phase 40 keV and 100 keV monochromatic image is 441.6 HU, and 100.2 HU, respectively. (C) Arterial phase NIC_A and NIC_P are 0.38, and 0.69, respectively. (D) Arterial phase Z_{eff} is 9.39. (E) US images. (F, G) CT value of venous phase phase 40 keV and 100 keV monochromatic image is 357.0 HU, and 95.8 HU, respectively. (H) Venous phase NIC_A and NIC_P are 0.69, and 0.73, respectively. (I) Venous phase Z_{eff} is 9.01. (J) Photomicrograph confirmed the pathological finding of the nodule as nodular goiter. (Hematoxylin-eosin stair; original magnification, 40). (K) Arterial phase λ_{HU} of the energy curve is 5.69 HU/keV. (L) Venous phase λ_{HU} of the energy curve is 4.35 HU/keV. HT, Hashimoto's thyroiditis; AP, arterial phase; VP, venous phase; HU, CT value; λ_{HU} , the slope of spectral HU curve; IC_{nodule}, iodine concentration of thyroid nodule; NIC_P. IC_{nodule}/IC_{thyroid} parenchyma; NIC_A IC_{nodule}/IC_{carotid} artery; λ_{HU} , slope of the spectral Hounsfield unit curve; Z_{eff}, effective atomic number; US, Ultrasound; H&E, Haematoxylin-eosin.

images, clinical indicators, genetic features, and pathological features to reduce the indeterminate diagnosis of patients with TNs, especially for organ donors who might contribute to the transmission of thyroid cancer (38).

Indeed, this study has some limitations. Firstly, it is a singlecenter retrospective study with a relatively small sample size. Despite perfect internal validation, the built DLCT-clinical nomogram is not yet suitable for general use before external validation of the predictive nomogram. Thus, multicenter and external prospective studies with larger sample sizes are needed to further validate the diagnostic power of the nomogram. Secondly, since the malignant TNs in our study were papillary thyroid carcinomas, the efficacy of our nomogram may be limited in identifying other pathological types of malignant TNs.

Conclusion

In conclusion, this study provides a DLCT-clinical nomogram, consisting of AP-NIC_P, HT, and BRAF V600E, which demonstrates favorable performance in detecting malignant lesions in cytologically indeterminate TNs. This validated nomogram can be an effective tool to assist clinicians in assessing the nature of cytologically indeterminate TN and reduce unbeneficial management.

Data availability statement

The raw data supporting the conclusions of this article will be made available by the authors, without undue reservation.

Ethics statement

The studies involving humans were approved by the Review Committee of Chongqing General Hospital. The studies were conducted in accordance with the local legislation and institutional requirements. Written informed consent for participation was not required from the participants or the participants' legal guardians/next of kin in accordance with the national legislation and institutional requirements.

Author contributions

XR: Writing – original draft, Writing – review & editing, Methodology. JZ: Data curation, Writing – original draft, Writing – review & editing. ZS: Methodology, Supervision, Visualization, Writing – original draft, Writing – review & editing. QL: Methodology, Writing – original draft, Writing – review & editing. DZh: Data curation, Funding acquisition, Writing – original draft, Writing – review & editing. XL: Data curation, Writing – original draft, Writing – review & editing. JY: Data curation, Methodology, Writing – original draft, Writing – review & editing. ZL: Data curation, Writing – original draft, Writing – review & editing. YW: Data curation, Writing – original draft, Writing – review & editing. DZe: Data curation, Writing – original draft, Writing – review & editing. DZe: Data curation, Writing – original draft, Writing – review & editing. DZe: Data curation, Writing – original draft, Writing – review & editing. DZe: Data curation, Writing – original draft, Writing – review & editing. XZ: Writing – original draft, Writing – review & editing. Conceptualization, Funding acquisition, Supervision, Writing – original draft, Writing – review & editing.

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Funding

The author(s) declare financial support was received for the research, authorship, and/or publication of this article. This study received funding from the medical research Key Program of the combination of the Chongqing National health commission and Chongqing Science and Technology bureau, China (NO.2019 ZDXM010), the medical research Program of the combination of Chongqing National health commission and Chongqing Science and Technology bureau, China (NO.2020FYYX151) and the Chongqing Postdoctoral Science Foundation (CSTB2022 NSCQ-BHX0737).

Conflict of interest

Author XZ was employed by Philips Healthcare.

The remaining authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: https://www.frontiersin.org/articles/10.3389/fonc.2024.1357419/ full#supplementary-material

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