

Research Article

Associations of Combined Exposure to Metabolic and Inflammatory Indicators with Thyroid Nodules in Adults: A Nested Case-Control Study

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Received 30 November 2023; Revised 9 February 2024; Accepted 19 February 2024; Published 27 March 2024

Academic Editor: Alexander Schreiber

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Objective. To explore associations of combined exposure to metabolic/inflammatory indicators with thyroid nodules. *Methods.* We reviewed personal data for health screenings from 2020 to 2021. A propensity score matching method was used to match 931 adults recently diagnosed with thyroid nodules in a 1:4 ratio based on age and gender. Conditional logistic regression and Bayesian kernel machine regression (BKMR) were used to explore the associations of single metabolic/inflammatory indicators and the mixture with thyroid nodules, respectively. *Results.* In the adjusted models, five indicators (OR_{Q4 vs. Q1}: 1.30, 95% CI: 1.07–1.58 for fasting blood glucose; OR_{Q4 vs. Q1}: 1.30, 95% CI: 1.08–1.57 for systolic blood pressure; OR_{Q4 vs. Q1}: 1.26, 95% CI: 1.04–1.53 for diastolic blood pressure; OR_{Q4 vs. Q1}: 1.23, 95% CI: 1.02–1.48 for white blood cell; OR_{Q4 vs. Q1}: 1.28, 95% CI: 1.07–1.55 for neutrophil) were positively associated with the risk of thyroid nodules, while high-density lipoproteins (OR_{Q3 vs. Q1}: 0.75, 95% CI: 0.61–0.91) were negatively associated with the risk of thyroid nodules. Univariate exposure-response functions from BKMR models showed similar results. Moreover, the metabolic and inflammatory mixture exhibited a significant positive association with thyroid nodules in a dose-response pattern, with systolic blood pressure being the greatest contributor within the mixture (conditional posterior inclusion probability of 0.82). No interaction effects were found among the five indicators. These associations were more prominent in males, participants with higher age (\geq 40 years old), and individuals with abnormal body mass index status. *Conclusions.* Levels of the metabolic and inflammatory mixture have a linear dose-response relationship with the risk of developing thyroid nodules, with systolic blood pressure levels being the most important contributor.

1. Introduction

Thyroid nodules are overgrown masses of normal thyroid cells in the gland [1], which are classified into several types: single, multiple, solid, or cystic [2]. A previous study reported that the global prevalence of thyroid nodules has reached 4–7%, of which 8–16% turn into thyroid cancer [3]. In recent years, the prevalence of thyroid nodules in China has shown a concerning increase, with various studies reporting prevalence rates ranging from 10% to 50% [4–7]. For example, a study involving 6,985,956 participants (mean age: 42.1 ± 13.1 years) from 30 provinces and regions in

China indicated an overall prevalence of thyroid nodules of 36.9% [4]. Although the rate of thyroid nodules evolving into thyroid cancer in China was similar to the findings of Burman and Wartofsky [3, 8], the high prevalence of thyroid nodules has made thyroid cancer the seventh most prevalent malignant tumor in China [9]. In addition, thyroid nodules may cause a variety of clinical sequelae such as thyroid dysfunction, dysphagia, and shortness of breath [10]. These findings suggest that thyroid nodules are a concern.

Clinical treatment of thyroid nodules is currently controversial [11]; thus, the prevention of thyroid nodules may be an effective strategy. For example, China launched a mandatory universal salt iodization program in 1996, which has been effective in controlling thyroid-related diseases [12]. However, recent studies also suggested that excessive iodine intake may increase the risk of thyroid nodules [12–14]. Thus, there is an urgent need to identify several additional modifiable risk factors, especially key factors, in adults.

Thyroid nodules are a common clinical disease caused by various factors [15]. Recent studies have shown that in addition to genetic, environmental exposure factors are more important for thyroid nodules [16]. Several environmental risk factors have been observed, including improved iodine intake [10, 17], exposure to various toxic compounds [18], and metabolic disorders and inflammatory responses [19, 20]. Metabolic disorders, including hypertension, dyslipidemia, and impaired fasting glucose, all contribute to increased thyroid nodules [21-26]. Inflammatory parameters such as white blood cell (WBC), neutrophil (N), lymphocyte (L), and monocyte (M) were significantly associated with thyroid nodules [27]. However, the exact pathophysiological pathways underlying the adverse effects of metabolic disorders and inflammatory responses on thyroid nodules are unclear. The main pathogenesis may be related to serum thyroid stimulating hormone (TSH) levels [28], which were positively associated with thyroid nodules [29]. Several studies have reported that blood pressure [30], fasting glucose [31], and inflammatory factors [8] were associated with TSH.

There were some limitations in previous studies. Firstly, most previous studies only focused on the single effects of metabolic/inflammation indicators on thyroid nodules [21, 22, 27]. To our knowledge, no study has explored the combined effects of multiple indicators. But humans often are exposed to multiple indicators simultaneously, which may have interaction effects on health [32]. Moreover, the available data are more based on cross-sectional studies and/ or limited sample sizes [33]. Thus, we employed a retrospective nested case-control study utilizing a relatively large sample to explore the relationships of combined exposure to metabolic and inflammatory indicators with thyroid nodules in this study. Combining previous articles and our database, we selected eight metabolic indicators and six inflammatory indicators that may be associated with thyroid nodules [21-27]. Our study aims to (1) assess the associations of the combined exposure to eight metabolic indicators and six inflammatory indicators with the occurrence of thyroid nodules, (2) explore the indicators that may have the greatest impact on thyroid nodules in the mixture, (3) investigate interactions among mixture components, and (4) discover the susceptible subgroups.

2. Methods

2.1. Study Population. The data reported in this study came from the Health Examination Center of Zhongda Hospital affiliated with Southeast University, which was performed in Nanjing City, Jiangsu Province, China, from January 1, 2020, to December 31, 2021. Clients with severe diseases or clinical symptoms were led to the emergency or outpatient

department. Thus, participants in this study were considered healthy or only have mild illnesses for physical examination. Inclusion criteria were (1) participants with health examination records in 2020 and 2021; (2) years ≥ 18 ; (3) there was no abnormal change in thyroid ultrasound in 2020; and (4) comprehensive examination of metabolic/inflammatory indicators in 2020. Exclusion criteria were (1) a history of thyroid surgery; (2) suspected Graves' disease or thyroid cancer; and (3) data missing on metabolic/inflammatory indicators in 2020. According to the inclusion and exclusion criteria, 931 adults recently diagnosed with thyroid nodules and 3724 controls (1:4 matched by age and gender using propensity score matching) were included in the final analysis (Figure S1). The principles of the Helsinki Declaration were followed. All data involving medical records were not publicly available, and no participants have been contacted. Ethical approval was obtained from the ethics committee of the Clinical Research Ethics Committee of Zhongda Hospital Affiliated with Southeast University (No.: 2022ZDSYLL218-P01).

2.2. Demographic and Anthropometric/Indicator Assessment. Demographic information was collected for the year of 2020, including gender, age, body mass index (BMI), smoking, drinking, diabetes, and hypertension. Anthropometric measurements were performed by a professionally trained nurse. Participants included in the study were required to take off their shoes, and then their height and weight were measured. In addition, participants were required to rest for at least 5-10 minutes before blood pressure measurement. Participants were asked to remain fasted from 10:00 pm the previous night and have their blood drawn by a nurse the next morning (8: 00-9: 30). A biochemical automatic analyzer (Dimension RxL Max, Siemens Corporation, German) was used to detect metabolic parameters. The whole blood count was detected by a whole blood automatic analyzer (BC-6800Plus, Mindray Medical, China).

2.3. Metabolic and Inflammatory Indicators. Combining previous articles and our database, we selected eight metabolic indicators and six inflammatory indicators of 2020. Metabolic indicators included total cholesterol (TC), low-density lipoprotein (LDL), triglyceride (TG), high-density lipoprotein (HDL), fasting blood glucose (FBG), uric acid (UA), systolic blood pressure (SBP), and diastolic blood pressure (DBP). Inflammatory parameters included WBC, M, basophils (B), eosinophils (E), L, and N.

2.4. Definition of Thyroid Nodules. Thyroid ultrasonography is performed by experienced sonographers using a highfrequency probe. Thyroid nodules have been defined as any solid (including solid with cystic component) and nodular lesion, which are different from the adjacent parenchyma in the thyroid gland by ultrasonography [15]. Laboratory technicians were trained by technical support staff to use the machines for analysis and to calibrate the analyzers according to standard quality assurance protocols. 2.5. Potential Covariates. Confounding factors were considered as gender (male and female), age (<30, 30–39, 40–49, and ≥50 years old), diabetes (no and yes), hypertension (no and yes), smoking (no and yes), drinking (no and yes), and BMI (BMI<18.5, 18.5–23.9, 24–27.9, and >28 kg/m²). Diabetes was identified by a FBG level of ≥7.0 mmol/L (126 mg/ dL). Hypertension was identified by a SBP of ≥140 mmHg or a DBP of ≥90 mmHg.

2.6. Statistical Analysis. The characteristics and metabolic/ inflammatory indicators across two groups (thyroid nodules and nonthyroid nodules) were compared using the chisquare test for categorical variables and *t*-tests for continuous variables. Spearman's correlation coefficient was used to assess the correlations between the indicators measured at the baseline. In addition, intraclass correlation coefficients (ICCs) were used to explore the correlations between indicators measured at 2020 and 2021. In the following regression models, we used a zero-mean normalization approach to standardize metabolic and inflammatory indicators. The histograms showed that most of the indicators after zero-mean normalization were normally or approximately normally distributed (Figure S2).

Multivariate conditional logistic regression models were employed to evaluate associations between multiple indicators and thyroid nodules. Participants were categorized into quartiles based on the level of each indicator, and the lowest quartile was used as a reference. Adjusted odds ratio (OR) and 95% confidence interval (CI) were calculated for the occurrence of thyroid nodules. The covariates included age, gender, diabetes, and hypertension. The model was adjusted for matching variables to account for residual confounding. False discovery rate (FDR) corrections were used to adjust *p* values. Since there were many indicators in this study, we screened five indicators (HDL, FBG, SBP, WBC, and N) based on the above regression results and correlation coefficients (Figure S3) and then included them in the subsequent BKMR models. Given the nonlinear and interactive effects, BKMR analysis was performed to assess the combined effects of multiple metabolic and inflammatory indicators [32]. The models were executed up to 10,000 iterations using a Markov chain Monte Carlo algorithm [34]. Five indicators were classified into two groups based on metabolic and inflammatory indicators. We selected the key indicators for thyroid nodules by calculating the group posterior inclusion probability (groupPIP) and conditional posterior inclusion probability (condPIP), where the threshold value of PIP was 0.5 [35]. The results of BKMR analysis were as follows: (1) nonlinear and/or nonadditive associations of individual indicators with the risk of thyroid nodules, (2) joint effects of the indicator mixture on the risk of thyroid nodules, (3) the relative importance of individual indicators within the mixture, and (4) interactive effects among mixture components.

Stratified analyses according to gender (male, female), age (<40, \geq 40 years), and BMI (18.5–23.9, <18.5/>23.9 kg/m²) were conducted. Furthermore, we conducted three sensitivity analyses to evaluate the stability of results. First,

given the possible confounding and mediating effects of BMI in these associations, BMI was not adjusted in the formal analysis. In the sensitivity analysis, we examined the potential confounding effect of BMI by adding the BMI variable to the BKMR model. Although smoking and drinking are risk factors for thyroid nodules, there is a large amount of missing data for these two factors in this study. We grouped the missing values of smoking or drinking into a category. In the second sensitivity analysis, we examined the potential confounding effect of two factors (dummy variables) by adding them to the BKMR model. Finally, all metabolic and inflammatory indicators were included in the BKMR model, and the covariates were controlled for age, gender, diabetes, and hypertension. SPSS (version 20; IBM SPSS Statistics) and R (version 4.0.2; R Foundation for Statistical Computing) were used to conduct statistical analysis. Two-sided P values below 0.05 were considered statistically significant.

3. Results

3.1. Characteristics of the Study Population. The characteristics of the participants are presented in Table 1. Of all participants, 61.2% were female and 57.3% were aged 40 years and above. A total of 931 individuals were newly diagnosed with thyroid nodules during the study period. The median maximum diameter of thyroid nodules was 0.30 (interquartile range: 0.24–0.40) mm (Figure S4). Comparisons of the risk of thyroid nodules across groups are also shown in Table 1. Participants with thyroid nodules were more likely to have higher BMI, diabetes, and hypertension than those with nonthyroid nodules. In addition, participants with thyroid nodules had higher levels of LDL, FBG, SBP, DBP, WBC, and N, whereas no significant differences were found for other indicators (Table 1).

Spearman's correlation coefficients between metabolic/ inflammatory indicators are displayed in Table S1. Spearman's correlation coefficients ranged from -0.010 to 0.890, with the highest correlation coefficient between TC and LDL (r = 0.888), and the remaining ones in descending order were SBP and DBP (r = 0.787) and WBC and M (r = 0.663). In addition, Table S2 showed that the correlations between all indicators for 2020 and 2021 were moderate to strong (ICCs: 0.520–0.839).

3.2. Single Indicator Exposure and Thyroid Nodules. The associations of metabolic and inflammatory indicators with thyroid nodules are shown in Table S3 and Table 2. After adjusting for age, gender, diabetes, and hypertension, the models showed that five indicators (FBG, SBP, DBP, WBC, and N) had significant positive associations with thyroid nodules (*P* value <0.05), but HDL had a significant negative association (*P* value <0.05). For example, compared with participants in the lowest quartile of FBG, SBP, DBP, WBC, and N, participants in the highest quartile showed 30% (95% CI: 1.07, 1.58), 30% (95% CI: 1.08, 1.57), 26% (95% CI: 1.04, 1.53), 23% (95% CI: 1.02, 1.48), and 28% (95% CI: 1.07, 1.55) increased the risk of thyroid nodules, respectively. Moreover, compared with participants in the lowest in the lowest quartile of the lowest quartile of the participants in the lowest quartile of the participants in the risk of thyroid nodules, respectively.

TABLE 1: Characteristics and metabolic/inflammatory indicators of the study population.

Variables	Total	Thuroid nodules	Nonthyroid nodules	v^2/t	D value
	10101			XT	1 value
fotal (%)	4055	931 (20.00)	3724 (80.00)	0.00	0.775
Gender (%)	1007 (20.00)	2(2, (20, 20)	1441 (70.90)	0.08	0.775
Male	1806 (38.80)	362 (20.20)	1441 (79.80)		
Female	2849 (61.20)	566 (19.90)	2283 (80.10)	2.02	0.560
Age (years, %)	722(15.50)	146 (20.20)	E76 (70.80)	2.02	0.569
< 30	1265 (15.50)	140(20.20)	5/6 (79.80)		
30-39	1265(27.20)	238 (18.80)	1027 (81.20)		
40-49	1234 (26.50)	246 (19.90)	988 (80.10)		
≥ 50	1434 (30.80)	301 (21.00)	1133 (79.00)	11 70	0.000
BMI (kg/m , %)	227 (4.00)	22(1450)	104 (05 50)	11./8	0.008
<18.5	227 (4.90)	33 (14.50)	194 (85.50)		
18.5-23.9	2525 (54.20)	477 (18.90)	2048 (81.10)		
24.0-27.9	1503 (32.30)	336 (22.40)	1167 (77.60)		
≥ 28.0	400 (8.60)	85 (21.30)	315 (78.80)		
Smoking (%) ^{ab}					
No	1292 (27.80)	245 (19.00)	1047 (81.00)	2.10	0.350
Yes	372 (8.00)	69 (18.50)	303 (81.50)		
Missing	2991 (64.30)	617 (20.60)	2374 (79.40)		
Drinking (%) ^{ab}					
No	1133 (24.30)	219 (19.30)	914 (80.70)	7.68	0.422
Yes	528 (11.30)	95 (18.00)	433 (82.00)		
Missing	2994 (63.20)	617 (20.60)	2377 (79.40)		
Diabetes (%) ^c				4.99	0.025
No	4504 (96.80)	890 (19.80)	3614 (80.20)		
Yes	151 (3.20)	41 (27.20)	110 (72.80)		
Hypertension (%) ^d				7.41	0.006
No	3591 (77.10)	687 (19.10)	2904 (80.90)		
Yes	1064 (22.90)	244 (22.90)	820 (77.10)		
Metabolic indicators					
TC (mmol/L)	4.82 ± 0.89	4.87 ± 0.94	4.81 ± 0.88	-1.81	0.070
LDL (mmol/L)	2.73 ± 0.67	2.77 ± 0.69	2.71 ± 0.66	-2.22	0.027
TG (mmol/L)	1.39 ± 0.94	1.44 ± 0.96	1.38 ± 0.96	-1.80	0.072
HDL (mmol/L)	1.44 ± 0.29	1.43 ± 0.29	1.45 ± 0.29	1.46	0.144
FBG (mmol/L)	5.26 ± 0.76	5.31 ± 0.77	5.24 ± 0.76	-2.50	0.013
UA (mmol/L)	332.80 ± 87.37	335.89 ± 87.73	332.03 ± 87.27	-1.20	0.228
SBP (mmHg)	125.06 ± 18.35	126.84 ± 18.84	124.61 ± 18.20	-3.31	0.001
DBP (mmHg)	75.67 ± 11.80	76.75 ± 11.90	75.40 ± 11.76	-3.13	0.002
Inflammatory indicators					
WBC (10 ⁹ cells/L)	6.08 ± 1.47	6.20 ± 1.46	6.05 ± 1.47	-2.65	0.008
M (10^9 cells/L)	0.35 ± 0.11	0.35 ± 0.11	0.34 ± 0.11	-1.88	0.060
B (10^9 cells/L)	0.03 ± 0.02	0.03 ± 0.02	0.03 ± 0.02	-1.26	0.270
$E (10^9 \text{ cells/L})$	0.14 ± 0.11	0.14 ± 0.11	0.14 ± 0.11	0.51	0.613
$L (10^9 \text{ cells/L})$	1.99 ± 0.57	2.02 ± 0.55	2.00 ± 0.57	-1.68	0.092
N (10^9 cells/L)	3.58 ± 1.12	3.66 ± 1.12	3.56 ± 1.12	-2.47	0.013

BMI: body mass index; TC: total cholesterol; LDL: low-density lipoprotein; TG: triglycerides; HDL: high-density lipoprotein; FBG: fasting blood glucose; UA: uric acid; SBP: systolic blood pressure; DBP: diastolic blood pressure; WBC: white blood cell; M: monocyte; B: basophil; E: eosinophil; L: lymphocyte; N: neutrophil. ^aDue to the missing covariate data, subgroup totals may not sum to the total sample population. ^bThe chi-square values were calculated without including missing values. ^cDiabetes was identified by a fasting blood glucose level of \geq 7.0 mmol/L (126 mg/dL). ^dHypertension was identified by a SBP of \geq 140 mmHg or a DBP of \geq 90 mmHg. Bold values indicate statistical significance, *P* < 0.05.

HDL, participants in the 3rd quartile showed a 25% (95% CI: 0.61, 0.91) decreased the risk of thyroid nodules. After FDR adjustments were made, similar results of statistically significant were found. Since there were many indicators in this study, we screened five indicators (HDL, FBG, SBP, WBC, and N) based on the above regression results and correlation coefficients (Figure S3) and then included them in the subsequent BKMR models.

3.3. BKMR Analyses. Figure 1 also showed linear relationships between exposure to single indicators and thyroid nodules when other indicators' exposure was fixed at the median. Figure 2 showed that a significant joint effect of the five indicators was found when all indicators were at or above their 55th percentile compared to the median. In addition, a slightly decreased and positive association between SBP and thyroid nodules was found when the other four indicators

	6 6	U		
Matabalic indicators		Adjusted models ^a		
Metabolic indicators	OR (95% CI)	P value	P value ^b	
TC				
01	1	_	_	
02	1.08 (0.90, 1.30) 0.413	0.707	
03	1.07 (0.89, 1.29) 0.471	0.707	
04	1 01 (0 84, 1 22) 0.883	0.883	
P-trend	0.956) 0.000	0.000	
	0.550			
LDL O1	1			
QI	1 12 (0 02 1 26) 0.244	0 452	
Q2	1.12 (0.92, 1.30)) 0.244	0.455	
Q3	1.11 (0.91, 1.34) 0.502	0.455	
Q4 D turn 1	1.06 (0.87, 1.28) 0.583	0.585	
P-trend	0.6/4			
TG				
Q1	1	. —	_	
Q2	0.99 (0.82, 1.20) 0.933	0.933	
Q3	1.10 (0.91, 1.34) 0.307	0.461	
Q4	1.15 (0.94, 1.40) 0.164	0.461	
<i>P</i> -trend	0.102			
HDL				
Q1	1	_	_	
Q2	0.83 (0.68, 1.00) 0.056	0.084	
Q3	0.75 (0.61, 0.91) 0.005	0.015	
Q4	0.87 (0.71, 1.06) 0.162	0.162	
P-trend	0.188			
FRG				
01	1	_	_	
Ω^1	1 10 (0 00 1 33) 0348	0 3/8	
Q2 03	1.10(0.90, 1.99) 1.20(1.00, 1.45)) 0.056	0.040	
Q3 04	1.20 (1.00, 1.43) 1.20 (1.07, 1.59)) 0.050	0.004	
Q4 D trand	1.50 (1.07, 1.58) 0.008	0.024	
<i>r</i> -tiend	0.003			
UA				
QI		\		
Q2	1.03 (0.85, 1.24) 0.795	0.795	
Q3	1.10 (0.89, 1.36) 0.363	0.626	
Q4	1.10 (0.88, 1.37) 0.417	0.626	
<i>P</i> -trend	0.359			
SBP				
Q1	1	_	—	
Q2	0.96 (0.79, 1.17) 0.692	0.865	
Q3	0.98 (0.81, 1.19) 0.865	0.865	
Q4	1.30 (1.08, 1.57) 0.006	0.018	
P-trend	0.004			
DBP				
01	1	_	_	
Õ2	0.99 (0.81, 1.20) 0.916	0.916	
03	1.09 (0.90, 1.31) 0.369	0.554	
Õ4	1.26 (1.04, 1.53) 0.016	0.048	
P-trend	0.009	,	0.010	
	5.007			

TABLE 2: Associations between metabolic/inflammatory indicators and thyroid nodules using conditional logistic regression.

bry indicators	OR (95% CI)	P value	P value ^b
	1	_	_
	1.03 (0.86, 1.25)	0.724	0.724
	1.23 (1.03, 1.48)	0.026	0.044
	1.23 (1.02, 1.48)	0.029	0.044
	0.095		
	1	_	_
	1.01 (0.84, 1.20)	0.955	0.955
	1.09 (0.89, 1.32)	0.406	0.609
	1.20 (0.99, 1.44)	0.062	0.186
	0.057		
	1	_	_
	1.16 (0.97, 1.39)	0.111	0.167

Adjusted models^a

Inflammatory indicators

WBC Q1 Q2 Q3 Q4 P-trend М Q1 Q2 Q3 Q4 P-trend L Q1 Q2 Q3 Q4 P-trend В Q1 Q2 Q3 Q4 P-trend Ε Q1 Q2 Q3 Q4 P-trend Ν Q1

Q2

1		
1.03 (0.86, 1.25)	0.724	0.724
1.23 (1.03, 1.48)	0.026	0.044
1.23 (1.02, 1.48)	0.029	0.044
0.095		
1	_	_
1.01 (0.84, 1.20)	0.955	0.955
1.09 (0.89, 1.32)	0.406	0.609
1.20 (0.99, 1.44)	0.062	0.186
0.057		
1	_	_
1.16 (0.97, 1.39)	0.111	0.167
1.01 (0.83, 1.21)	0.957	0.957
1.19 (0.99, 1.43)	0.061	0.167
0.625		
1	_	_
0.95 (0.79, 1.14)	0.564	0.564
1.17 (0.97, 1.40)	0.104	0.312
1.08 (0.89, 1.30)	0.438	0.564
0.256		
1	_	_
1.07 (0.89, 1.30)	0.463	0.753
1.06 (0.87, 1.29)	0.547	0.753
0.97 (0.79, 1.18)	0.753	0.753
0.311		

Q3 0.020 0.030 1.25 (1.04, 1.50) 1.28 (1.07, 1.55) 0.024 Q4 0.008 P-trend 0.040 TC: total cholesterol; LDL: low-density lipoprotein; TG: triglycerides; HDL: high-density lipoprotein; FBG: fasting blood glucose; UA: uric acid; SBP: systolic blood pressure; DBP: diastolic blood pressure; WBC: white blood cell; M: monocyte; B: basophil; E: eosinophil; L: lymphocyte; N: neutrophil. ^aAdjusting age, gender, diabetes, and hypertension. ^bFalse discovery rate (Benjamini and Hochberg). Bold values indicate statistical significance, P < 0.05.

1 1.04 (0.86, 1.26)

0.683

0.683

were fixed at different percentiles (25th, 50th, or 75th) (Figure 3). SBP exhibited strong linear associations, which was supported by PIPs in Table 3. One groupPIPs were higher than 0.5, and the condPIP of SBP (0.82) was the highest in the group. Finally, we estimated bivariate exposure-response functions for the five indicators (Figure 4). We did not find a significant interaction effect among the five indicators.



FIGURE 1: Univariate exposure-response functions and 95% confidence intervals for associations between single metabolic/inflammatory indicators and the risk of thyroid nodules when other indicators were fixed at the median. Adjusted variables included age, gender, diabetes, and hypertension.



FIGURE 2: The overall effect of the mixture (95% confidence intervals) when all of the indicators were fixed at a specific quantile (ranging from the 25th percentile to 75th percentile), as compared to when all indicators were fixed at their median values (the 50th percentile). Adjusted variables included age, gender, diabetes, and hypertension.

3.4. Subgroup and Sensitivity Analysis. The stratified analyses by gender, age, and BMI showed that these joint associations were more obvious in males, participants with higher age (\geq 40 years old), and individuals with abnormal BMI (Figures S5–S10). Besides, after controlling for BMI (continuous), smoking and drinking, or all 14 indicators, three sensitivity analyses did not materially change our findings (Figures S11–S13).

4. Discussion

4.1. Key Findings. In the adjusted model, FBG, SBP, DBP, WBC, and N were significantly positively correlated with thyroid nodules compared to their lowest concentration groups, while HDL was significantly negatively correlated. Univariate exposure-response functions from BKMR models



FIGURE 3: Single-exposure effects (95% confidence intervals), defined as the changes in the risk of thyroid nodules associated with a change in a particular indicator from its 25th to its 75th percentile, where all of the remaining indicators were fixed at a specific quantile (the 25th, 50th, or 75th percentile). Adjusted variables included age, gender, diabetes, and hypertension.

showed similar results. Moreover, our study found a linear dose-response relationship between the mixture of metabolic and inflammatory indicators and thyroid nodules, and SBP was the most important contributor within the mixture. Nevertheless, no interaction effects were found among the five indicators. These associations were more prominent in males, participants with higher age (≥ 40 years old), and individuals with abnormal BMI. To our knowledge, this is the first study to examine associations of combined exposure to metabolic and inflammatory indicators with thyroid nodules.

4.2. Metabolic and Inflammatory Indicators. Consistent with the results of most previous studies [4, 33], our study also showed that blood pressure was positively associated with

Variables	Group	groupPIP	condPIP
HDL	1	0.84	0.02
FBG	1	0.84	0.16
SBP	1	0.84	0.82
Ν	2	0.48	0.54
WBC	2	0.48	0.46

TABLE 3: PIPs for group inclusion and conditional inclusion into thyroid nodules using Bayesian kernel machine regression (BKMR) models.

PIP: Posterior inclusion probabilities; HDL: high-density lipoprotein; FBG: fasting blood glucose; SBP: systolic blood pressure; WBC: white blood cell; N: neutrophil.



FIGURE 4: Bivariate exposure response functions. Each cell represented the exposure-response curve for the column indicator when the row indicator was fixed at 10th, 50th, and 90th percentiles and the remaining indicators were fixed at their medians.

thyroid nodules. A recent meta-analysis [33] showed that abnormal blood pressure was associated with thyroid nodules (OR = 1.68, 95% CI: 1.62-1.75). Several large-scale studies, such as [36, 37], have concluded that abnormal blood pressure is a risk factor for thyroid nodules. Other studies [26, 38] with small samples have yielded inconsistent results. The exact mechanism of the risk of thyroid nodules due to hypertension is not known. Several studies have shown a positive correlation between TSH and SBP/DBP [39, 40], and high TSH levels in hypertensive patients may contribute to the formation of thyroid nodules. In addition, we cannot ignore the possibility of the potential confounding effect of TSH on the association between blood pressure and thyroid nodules. Unfortunately, only 30% of the individuals in this study underwent TSH measurement. This severe selection bias hindered the possibility of meaningful mediation analyses, and thus, further studies are warranted to clarify the underlying biological mechanisms.

Consistent with most previous studies [33, 41-44], we found a significant positive association between blood glucose and thyroid nodules. Recently, a meta-analysis [33] also showed that hyperglycemia was associated with thyroid nodules (OR = 1.59, 95% CI: 1.46-1.74). One possible explanation is the confounding effects of insulin resistance. On the one hand, some studies have shown that insulin resistance can promote the formation and growth of thyroid nodules [33, 45]; on the other hand, insulin resistance is a key factor in the pathogenesis of impaired glucose metabolism [46]. Regrettably, no data on insulin resistance were collected in this study, so the confounding effects of insulin resistance could not be ruled out. As we know, high levels of TSH can lead to the development of thyroid nodules [47, 48]. A study [41] has shown higher serum TSH levels in serum type 2 diabetic patients than in control prediabetic and control patients, providing another possible explanation.

In this study, higher levels of HDL were significantly negatively associated with thyroid nodules, which was consistent with the limited studies [4, 26]. A cross-sectional study showed that elevated HDL levels were negatively correlated with thyroid nodules, while TG and LDL were positively correlated [4]. Another case-control study also showed a significant association between low HDL (OR = 2.77, 95% CI: 1.44–5.30) and thyroid nodules. Compared to the previous two studies [4, 26], we used a retrospective nested case-control to provide relatively reliable evidence. However, the underlying mechanism by which high levels of serum HDL reduce the development of thyroid nodules remains unclear. Further studies are needed to examine the prospective association of HDL with thyroid nodules and to better understand the mechanisms.

Our retrospective nested case-control study showed that WBC and N increased the risk of thyroid nodules. Li et al. discovered a higher prevalence of thyroid nodules in participants with high levels of inflammation (WBC, N, L, and M) by using propensity score matching for metabolic parameters and other confounding factors [8]. Moreover, a retrospective cohort study (included 6587 participants) showed that M was a risk factor for thyroid nodules [48]. Haider et al. also reported that the monocyte-to-high-density lipoprotein cholesterol ratio (MHR) and neutrophil-to-lymphocyte ratio (NLR) were significantly associated with the presence of thyroid nodules [49]. These findings are consistent with our expectations. As we know, chronic inflammation plays a role in the development of thyroid nodules [50].

4.3. Combined Exposure. Considering the high correlation and complexity between indicators, traditional methods may not provide a true view of the relationship between mixed exposures to multiple indicators and thyroid nodules. However, to our knowledge, no epidemiological studies are addressing this issue. In this study, we used the BKMR model to assess associations of combined exposure to multiple indicators with thyroid nodules. First, consistent with our expectations, we found a linear dose-response relationship between the combined five metabolic/inflammatory indicators (HDL, FBG, SBP, WBC, and N) and the risk of thyroid nodules. Although both metabolic and inflammatory indicators are thought to have an impact on thyroid nodules, their relative importance remains unclear. In this study, we found that metabolic indicators were more important than inflammatory indicators. Within this mixture, blood pressure is the most important component. Our findings suggest that controlling metabolic indicators, especially blood pressure, may be important in reducing the risk of thyroid nodules in adults.

4.4. Subgroup Analysis. Identifying susceptible populations is important for both public health and clinical practice; however, knowledge in this area remains unclear. Most studies have shown a higher prevalence of thyroid nodules in females than in males [4, 22, 51], but few studies have explored the gender-specific associations between metabolic/ inflammatory indicators and thyroid nodules [22]. We

found that the association of joint exposure was more prominent in males. In addition, we also found that blood pressure was the main determinant in males and blood glucose in females. Ding et al. found that diabetes (OR = 1.47, 95% CI: 1.17-1.84) remained strongly and independently associated with a higher risk of thyroid nodules in females but not in males. In a retrospective cohort study, Huang et al. found that the association between the metabolic indicator (uric acid) and thyroid nodules was more pronounced in females [48]. Although limited results of the gender-specific association between metabolic indicators and thyroid nodules are controversial, all these results supported the fact that gender may play a moderating role in this association. It is still too early to draw any conclusions on the gender difference in the relationship between metabolic indicators and thyroid nodules, and further studies are warranted. Interestingly, we also found that associations of joint exposure were more prominent in participants with higher age (≥40 years old) and individuals with abnormal body mass index status, which suggested potential harmful effects of high age, under- or overweight status.

4.5. Implications for Public Health. Our findings may have implications for public health. The linear dose-response relationship of the mixture of metabolic and inflammatory indicators with thyroid nodules provides valuable insights. Notably, our findings highlight the dominant role of SBP in the mixture. This provides key clues for customizing prevention strategies, suggesting that focusing on managing SBP and considering a combination of approaches (e.g., appropriate medication use and exercise interventions) to intervene with metabolic and inflammatory factors. Importantly, our study reveals population-specific patterns in the observed associations. Males, individuals aged 40 years and older, and those with abnormal levels of BMI showed stronger associations, providing targeted information to optimize prevention strategies. This detailed understanding allows for the development of more targeted interventions for different populations. For example, for males and individuals with higher age, we can emphasize the critical role of SBP and encourage regular blood pressure monitoring and active blood pressure management. Meanwhile, for individuals with abnormal BMI, we recommend weight management and nutritional education programs to help keep their metabolism and inflammation in balance. These specific measures are expected to increase public awareness of health and motivate more people to adopt active lifestyles, thereby reducing the risk of thyroid nodules.

4.6. Limitations and Strengths. There are three strengths of this study. First, to our knowledge, this is the first study to examine the associations of combined exposure to metabolic/inflammatory indicators with thyroid nodules. Second, utilizing nested case-control studies helps mitigate selection bias, recall bias, and confounding bias, thereby enhancing the internal validity of associations. Third, we performed a series of subgroup and sensitivity analyses to show that the results were considerably robust.

However, this study also has several limitations. First, the study population was from one health examination center, which led to possible limitations in the generalization of our findings to other regions or the general population. Future studies could include data from multiple medical centers or different population characteristics to further validate our results. Second, it is important to note that because of the observational research design, we can only infer correlation, not causation. Third, one-time sample measures may bias internal exposure estimates. We explored the association between baseline and follow-up indicators and found moderate to strong reproducibility for these indicators (ICCs ranged from 0.520 to 0.839). Thus, we believe that one-time sample measures may reflect the long-term exposure levels to a certain extent. Fourth, we followed up for only one year, which was unlikely to affect our overall conclusions but limited our ability to assess different thyroid grades. For example, the median maximum diameter of thyroid nodules in this study was only 0.30 (interquartile range: 0.24-0.40) mm. Fifth, general several inflammatory indicators (e.g., Creactive protein and interleukin-6) were not measured. Therefore, these clinical indicators could not be considered in this analysis, which may underestimate the association of combined exposure to inflammatory indicators with thyroid nodules. Finally, since our data came from the health examination center, some confounding factors were not collected well. For example, there was a large amount of missing smoking and drinking, and we could only adjust for these factors in the sensitivity analysis. In addition, residual confounding of unmeasured variables (e.g., physical activities, dietary structure, and iodine content) cannot be excluded.

5. Conclusions

Our study found a linear dose-response relationship between the mixture of metabolic/inflammatory indicators and thyroid nodules, and SBP was the most important contributor within the mixture. These associations were more prominent in males, participants with higher age (\geq 40 years old), and individuals with abnormal BMI. Our findings suggest that reduced metabolic and inflammatory levels, especially reduced blood pressure levels, may be important in preventing thyroid nodules. Further studies are needed to explore the prospective association between metabolic/inflammatory indicators and thyroid nodules and to elucidate the complex mechanisms between these indicators and thyroid nodules.

Data Availability

The datasets used and/or analyzed during the current study are available from the corresponding author on reasonable request. All data involving medical records were not publicly available, and no participants have been contacted.

Ethical Approval

Ethical approval was obtained from the ethics committee of the Clinical Research Ethics Committee of Zhongda Hospital Affiliated with Southeast University (No.: 2022ZDSYLL218-P01).

Conflicts of Interest

The authors declare that they have no conflicts of interest.

Authors' Contributions

Xin-Yi Zhu conducted writing of the original draft, formal analysis, validation. Xing-Chen Meng conducted formal analysis and validation. Bei-Jing Cheng conducted formal analysis and validation. Chun Wang conducted investigation, data curation and validation. Jia Wang conducted investigation, data curation, and validation. Tian-Lin Li conducted investigation, data curation, and validation. Hui Li conducted investigation, data curation, and validation. Ke Meng conducted investigation, data curation, and validation. Ran Liu conducted investigation, data curation, writing of the review, editing, and validation.

Acknowledgments

The authors are grateful to the research group of the Health Examination Center of Zhongda Hospital affiliated with Southeast University. This work was supported by the National Natural Science Foundation of China (grant numbers: 82173479 and 81872579) and the Postgraduate Research & Practice Inovation Program of Jiangsu Province (grant number: KYCX230333).

Supplementary Materials

Supplementary 1. Table S1: correlations between metabolic/ inflammatory indicators using Spearman's correlation analysis in 2020. Supplementary 2. Table S2: intraclass correlation coefficients of metabolic/inflammatory indicators in 2020 and 2021. Supplementary 3. Table S3: associations between metabolic/inflammatory indicators and thyroid nodules using conditional logistic regression. Supplementary 4. Figure S1: flow chart. Supplementary 5. Figure S2: the histograms of metabolic/inflammatory indicators. Supplementary 6. Figure S3: correlations between metabolic/ inflammatory indicators using Spearman's correlation analysis. Supplementary 7. Figure S4: maximum diameter of thyroid nodules (mm) in 931 participants. Supplementary 8. Figure S5: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in male, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included age, diabetes, and hypertension. Supplementary 9. Figure S6: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in female, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included age, diabetes, and hypertension. Supplementary 10. Figure S7: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in participants with lower age (<40 years old), estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, diabetes, and hypertension. Supplementary 11. Figure S8: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in participants

with lower age (\geq 40 years old), estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, diabetes, and hypertension. Supplementary 12. Figure S9: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in participants with normal body mass index status $(18.5-23.9 \text{ kg/m}^2)$, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, age, diabetes, and hypertension. Supplementary 13. Figure S10: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in participants with abnormal body mass index status (<18.5 or >23.9 kg/m²), estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, age, diabetes, and hypertension. Supplementary 14. Figure S11: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in adults, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, age, diabetes, hypertension, and BMI (continuous). Supplementary 15. Figure S12: associations between five metabolic/inflammatory indicators and the risk of thyroid nodules in adults, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, age, diabetes, hypertension, smoking, and drinking. Supplementary 16. Figure S13: associations between 14 metabolic/inflammatory indicators and the risk of thyroid nodules in adults, estimated using Bayesian kernel machine regression (BKMR). Adjusted variables included gender, age, diabetes, and hypertension. (Supplementary Materials)

References

- B. Zou, L. Sun, X. Wang, and Z. Chen, "The prevalence of single and multiple thyroid nodules and its association with metabolic diseases in Chinese: a cross-sectional study," *International Journal of Endocrinology*, vol. 2020, Article ID 5381012, 11 pages, 2020.
- [2] T. G. Pemayun, "Current diagnosis and management of thyroid nodules," Acta Med Indones, vol. 48, no. 3, pp. 247– 257, 2016.
- [3] K. D. Burman and L. Wartofsky, "Thyroid nodules," New England Journal of Medicine, vol. 373, no. 24, pp. 2347–2356, 2015.
- [4] Y. Li, C. Jin, J. Li et al., "Prevalence of thyroid nodules in China: a health examination cohort-based study," *Frontiers in Endocrinology*, vol. 12, Article ID 676144, 2021.
- [5] M. Y. Qu, W. Tang, X. Y. Cui et al., "Increased prevalence of thyroid nodules across nearly 10 Years in Shanghai, China," *Current Medical Science*, vol. 43, no. 1, pp. 191–197, 2023.
- [6] L. Xu, F. Zeng, Y. Wang, Y. Bai, X. Shan, and L. Kong, "Prevalence and associated metabolic factors for thyroid nodules: a cross-sectional study in Southwest of China with more than 120 thousand populations," *Brihanmumbai Municipal Corporation Endocrine Disorders*, vol. 21, no. 1, p. 175, 2021.
- [7] W. Zhao, C. Han, X. Shi et al., "Prevalence of goiter and thyroid nodules before and after implementation of the universal salt iodization program in mainland China from 1985 to 2014: a systematic review and meta-analysis," *PLoS One*, vol. 9, no. 10, Article ID 109549, 2014.

- [8] Z. Li, Y. Huang, X. Chen, C. Wei, P. Yang, and W. Xu, "The effect of inflammation on the formation of thyroid nodules," *International journal of endocrinology*, vol. 2020, Article ID 9827349, 9 pages, 2020.
- [9] M. Cao, H. Li, D. Sun, and W. Chen, "Cancer burden of major cancers in China: a need for sustainable actions," *Cancer Communications*, vol. 40, no. 5, pp. 205–210, 2020.
- [10] Y. Zhu, M. Tong, Y. Wang et al., "Prevalence of thyroid nodules and its association with water iodine among Chinese men and women," *Environmental Research*, vol. 212, Article ID 113270, 2022.
- [11] E. J. Ha, S. R. Chung, D. G. Na et al., "2021 Korean thyroid imaging reporting and data system and imaging-based management of thyroid nodules: Korean society of thyroid radiology consensus statement and recommendations," *Korean Journal of Radiology*, vol. 22, no. 12, pp. 2094–2123, 2021.
- [12] X. Sun, Z. Shan, and W. Teng, "Effects of increased iodine intake on thyroid disorders," *Endocrinol Metab (Seoul)*, vol. 29, no. 3, pp. 240–247, 2014.
- [13] Y. Wang, J. Wang, Z. Chen et al., "Analysis of the correlation between high iodized salt intake and the risk of thyroid nodules: a large retrospective study," *Brihanmumbai Municipal Corporation Cancer*, vol. 21, no. 1, p. 1000, 2021.
- [14] D. E. Yan, L. Hu, Y. F. Shen et al., "Iodine status and its association with prevalence of thyroid diseases in adults from Jiangxi Province, China," *Endocrine*, vol. 82, no. 2, pp. 335– 342, 2023.
- [15] B. R. Haugen, E. K. Alexander, K. C. Bible et al., "2015 American thyroid association management guidelines for adult patients with thyroid nodules and differentiated thyroid cancer: the American thyroid association guidelines task force on thyroid nodules and differentiated thyroid cancer," *Thyroid*, vol. 26, no. 1, pp. 1–133, 2016.
- [16] C. Durante, G. Costante, G. Lucisano et al., "The natural history of benign thyroid nodules," *Japan Automobile Man*ufacturers Association, vol. 313, no. 9, pp. 926–935, 2015.
- [17] P. Laurberg, C. Cerqueira, L. Ovesen et al., "Iodine intake as a determinant of thyroid disorders in populations," *Best Practice & Research Clinical Endocrinology & Metabolism*, vol. 24, no. 1, pp. 13-27, 2010.
- [18] R. Zamora-Ros, V. Cayssials, S. Franceschi et al., "Polyphenol intake and differentiated thyroid cancer risk in the European Prospective Investigation into Cancer and Nutrition (EPIC) cohort," *International Journal of Cancer*, vol. 146, no. 7, pp. 1841–1850, 2020.
- [19] Y. Deng, J. Zhang, G. Zou et al., "Peripheral blood inflammatory markers can predict benign and malignant thyroid nodules," *International Journal of Endocrinology*, vol. 2022, Article ID 2319660, 6 pages, 2022.
- [20] J. Shin, M. H. Kim, K. H. Yoon, M. I. Kang, B. Y. Cha, and D. J. Lim, "Relationship between metabolic syndrome and thyroid nodules in healthy Koreans," *Korean Journal of Internal Medicine (English Edition)*, vol. 31, no. 1, pp. 98–105, 2015.
- [21] Y. Chen, C. Zhu, Y. Chen et al., "The association of thyroid nodules with metabolic status: a cross-sectional spect-China study," *International Journal of Endocrinology*, vol. 2018, Article ID 6853617, 8 pages, 2018.
- [22] X. Ding, Y. Xu, Y. Wang et al., "Gender disparity in the relationship between prevalence of thyroid nodules and metabolic syndrome components: the SHDC-CDPC community-based study," *Mediators of Inflammation*, vol. 2017, Article ID 8481049, 11 pages, 2017.

- [23] W. Guo, L. Tan, W. Chen et al., "Relationship between metabolic syndrome and thyroid nodules and thyroid volume in an adult population," *Endocrine*, vol. 65, no. 2, pp. 357–364, 2019.
- [24] L. Hu, T. Li, X. L. Yin, and Y. Zou, "An analysis of the correlation between thyroid nodules and metabolic syndrome," *Endocrine Connections*, vol. 9, no. 9, pp. 933–938, 2020.
- [25] S. Kir, Y. Aydin, and H. Coskun, "Relationship between metabolic syndrome and nodular thyroid diseases," *Scandinavian Journal of Clinical and Laboratory Investigation*, vol. 78, no. 1-2, pp. 6–10, 2018.
- [26] R. A. Mayers, A. Soria Montoya, A. Piscoya Rivera, and W. G. Silva Caso, "Association between metabolic syndrome and euthyroid nodular goiter: a case-control study," *Colombia Médica*, vol. 50, no. 4, pp. 239–251, 2019.
- [27] Y. Li, D. Teng, J. Ba et al., "Efficacy and safety of long-term universal salt iodization on thyroid disorders: epidemiological evidence from 31 provinces of mainland China," *Thyroid*, vol. 30, no. 4, pp. 568–579, 2020.
- [28] H. K. Chiu, S. Sanda, P. Y. Fechner, and C. Pihoker, "Correlation of TSH with the risk of paediatric thyroid carcinoma," *Clinical Endocrinology*, vol. 77, no. 2, pp. 316–322, 2012.
- [29] J. Rezzonico, M. Rezzonico, E. Pusiol, F. Pitoia, and H. Niepomniszcze, "Introducing the thyroid gland as another victim of the insulin resistance syndrome," *Thyroid*, vol. 18, no. 4, pp. 461–464, 2008.
- [30] F. Turchi, V. Ronconi, V. D. Tizio, M. Boscaro, and G. Giacchetti, "Blood pressure, thyroid-stimulating hormone, and thyroid disease prevalence in primary aldosteronism and essential hypertension," *American Journal of Hypertension*, vol. 24, no. 12, pp. 1274–1279, 2011.
- [31] M. J. Hu, C. Zhang, L. Liang et al., "Fasting serum glucose, thyroid-stimulating hormone, and thyroid hormones and risk of papillary thyroid cancer: a case-control study," *Head & Neck*, vol. 41, no. 7, pp. 2277–2284, 2019.
- [32] J. F. Bobb, B. Claus Henn, L. Valeri, and B. A. Coull, "Statistical software for analyzing the health effects of multiple concurrent exposures via Bayesian kernel machine regression," *Environmental Health*, vol. 17, no. 1, p. 67, 2018.
- [33] F. Zhang, Y. Li, X. Yu et al., "The relationship and gender disparity between thyroid nodules and metabolic syndrome components based on a recent nationwide cross-sectional study and meta-analysis," *Frontiers in Endocrinology*, vol. 12, Article ID 736972, 2021.
- [34] Y. Zhang, T. Dong, W. Hu et al., "Association between exposure to a mixture of phenols, pesticides, and phthalates and obesity: comparison of three statistical models," *Environment International*, vol. 123, pp. 325–336, 2019.
- [35] E. Coker, J. Chevrier, S. Rauch et al., "Association between prenatal exposure to multiple insecticides and child body weight and body composition in the VHEMBE South African birth cohort," *Environment International*, vol. 113, pp. 122– 132, 2018.
- [36] X. Lai, P. Ouyang, H. Zhu et al., "[Detection rate of thyroid nodules in routine health check-up and its influencing factors: a 10-year survey of 309 576 cases]," *Nan Fang Yi Ke Da Xue Xue Bao*, vol. 40, no. 2, pp. 268–273, 2020.
- [37] J. H. Moon, M. K. Hyun, J. Y. Lee et al., "Prevalence of thyroid nodules and their associated clinical parameters: a large-scale, multicenter-based health checkup study," *Korean Journal of Internal Medicine (English Edition)*, vol. 33, no. 4, pp. 753– 762, 2018.

- [38] D. Rendina, G. De Filippo, G. Mossetti et al., "Relationship between metabolic syndrome and multinodular non-toxic goiter in an inpatient population from a geographic area with moderate iodine deficiency," *Journal of Endocrinological Investigation*, vol. 35, no. 4, pp. 407–412, 2012.
- [39] S. B. Park, H. C. Choi, and N. S. Joo, "The relation of thyroid function to components of the metabolic syndrome in Korean men and women," *Journal of Korean Medical Science*, vol. 26, no. 4, pp. 540–545, 2011.
- [40] O. Topaloglu, F. Gokay, K. Kucukler et al., "Is autoimmune thyroiditis a risk factor for early atherosclerosis in premenopausal women even if in euthyroid status?" *Endocrine*, vol. 44, no. 1, pp. 145–151, 2013.
- [41] C. Anil, A. Akkurt, S. Ayturk, A. Kut, and A. Gursoy, "Impaired glucose metabolism is a risk factor for increased thyroid volume and nodule prevalence in a mild-to-moderate iodine deficient area," *Metabolism*, vol. 62, no. 7, pp. 970–975, 2013.
- [42] W. W. Dong, D. L. Zhang, Z. H. Wang, C. Z. Lv, P. Zhang, and H. Zhang, "Different types of diabetes mellitus and risk of thyroid cancer: a meta-analysis of cohort studies," *Frontiers in Endocrinology*, vol. 13, Article ID 971213, 2022.
- [43] H. Li and J. Qian, "Association of diabetes mellitus with thyroid cancer risk: a meta-analysis of cohort studies," *Medicine (Baltimore)*, vol. 96, no. 47, Article ID 8230, 2017.
- [44] Y. Yeo, S. H. Ma, Y. Hwang et al., "Diabetes mellitus and risk of thyroid cancer: a meta-analysis," *PLoS One*, vol. 9, no. 6, Article ID 98135, 2014.
- [45] Z. Heidari, M. A. Mashhadi, and S. Nosratzehi, "Insulin resistance in patients with benign thyroid nodules," *Archives of Iranian Medicine*, vol. 18, no. 9, pp. 572–576, 2015.
- [46] W. F. Rayburn, "Diagnosis and classification of diabetes mellitus: highlights from the American Diabetes Association," *Journal of Reproductive Medicine*, vol. 42, no. 9, pp. 585-586, 1997.
- [47] E. Ayroldi, M. G. Petrillo, M. C. Marchetti et al., "Long glucocorticoid-induced leucine zipper regulates human thyroid cancer cell proliferation," *Cell Death & Disease*, vol. 9, no. 3, p. 305, 2018.
- [48] Y. Huang, Z. Li, K. Yang et al., "The association of uric acid with the development of thyroid nodules: a retrospective cohort study," *BMC Endocrine Disorders*, vol. 22, no. 1, p. 197, 2022.
- [49] N. Haider, Z. Mahmood, F. Khalid, and S. A. Razzak, "Neutrophils to lymphocytes ratio between benign and malignant thyroid nodule," *Pakistan Journal of Medical Sciences*, vol. 37, no. 7, pp. 1908–1911, 2021.
- [50] S. Destek, B. Benturk, Y. Yapalak, and O. F. Ozer, "Clinical significance of erythrocyte sedimentation rate, leukocyte, fibrinogen, C-reactive protein, and pentraxin 3 values in thyroid nodules," *Sisli Etfal Hastan Tip Bul*, vol. 56, no. 2, pp. 270–275, 2022.
- [51] H. Guo, M. Sun, W. He et al., "The prevalence of thyroid nodules and its relationship with metabolic parameters in a Chinese community-based population aged over 40 years," *Endocrine*, vol. 45, no. 2, pp. 230–235, 2014.