Supplementary Online Content

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eTable 1. Basic Information of all Radiologists

eMethods.

eReference.

eTable 2. The Demographic Characteristics of Patients and the Nodules Distribution of ACR TI-RADS Category

eTable 3. Distribution of Images Features

eTable 4. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Composition Features in Junior and Senior Radiologists

eTable 5. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Echogenicity Features in Junior and Senior Radiologists

eTable 6. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Shape Features in Junior and Senior Radiologists

eTable 7. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Margin Features in Junior and Senior Radiologists

eTable 8. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Echogenic Foci Features in Junior and Senior Radiologists

eTable 9. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different Nodule Sizes in Junior and Senior Radiologists **eTable 10.** Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different Parenchymal Backgrounds in Junior and Senior Radiologists

eFigure 1. The Significant Features for Junior Radiologists

eFigure 2. The Significant Features for Senior Radiologists

This supplementary material has been provided by the authors to give readers additional information about their work.

	Radiol	ogist	years	Work Seniority (as radiologists, year)	Degree	Average examinations performed per month
	Junior	Reader 1	29	1.5	PhD	900
	radiologists	Reader 2	25	1.5	MM	800
		Reader 3	34	2	MM	1200
		Reader 4	31	2	PhD	800
Retrospective set		Reader 5	32	3	MD	600
	Senior	Reader 6	37	8	MM	780
	radiologists	Reader 7	39	9	PhD	800
		Reader 8	34	8	BMS	850
		Reader 9	38	10	PhD	1100
		Reader 10	37	12	MM	850
	Junior	Reader 11	29	1.5	MD	650
	radiologists	Reader 12	29	1.5	MM	780
Prospective set		Reader 13	30	2	MM	800
	Senior	Reader 14	37	10	MM	830
	radiologists	Reader 15	33	6	MM	1050
		Reader 16	37	10	PhD	1180

eTable 1. Basic Information of all Radiologists

Abbreviations: PhD, Doctor of Philosophy; MM, Master of Medicine; MD, Doctor of Medicine; BMS, Bachelor of Medical Science.

eMethods.

(1) The definitions of junior and senior radiologists are as follows:

Junior radiologists are doctors who have less than 3 years of experience in performing ultrasound scans. They undergo a 3-year standardized national resident training program, which includes standardized instruction in conducting thyroid ultrasound examinations. On average, they evaluate thyroid ultrasound images from approximately 600 patients per year. Senior radiologists, on the other hand, have more than 5 years of experience in thyroid ultrasound and have completed both the standardized national resident training and specialist training in ultrasound. They review thyroid ultrasound images from an average of about 800 patients per year.

(2) AI model establishment and evaluation

The deep-learning AI model was established in previous study¹. For the training set, a total of 18049 ultrasound images from 8339 patients with thyroid nodules were retrospectively retrieved from the individual thyroid imaging database at the First Affiliated Hospital of Sun Yat-sen University and Sun Yat-sen University Cancer Center, Guangzhou, China. AI deep-learning algorithm was specifically designed to diagnose malignancy from benign thyroid nodules, with a combined architecture of three networks: ResNet, ResNeXt, and DenseNet. To search for the optimal weights for each network branch and get the ensembled output, we used the brute-force search method via cross-test in the training sets. The final weighting ratios are 0.40 for ResNet, 0.35 for ResNeXt, and 0.25 for DenseNet. Then three independent testing sets were used to evaluate the AI model. In test A, the diagnostic performance of stand-alone AI was compared with radiologists. The results showed that AI achieved significantly higher accurate diagnosis (AUROC: 0.922, [95% CI 0.910-0.934]) than radiologists (AUROC: 0.839, [95% CI 0.834-0.844]; P < .0001). In test B, improvement in the diagnostic performance of -AI-assisted strategy was evaluated. AI assistance improved the pooled AUROC of the radiologists from 0.855 (95% CI 0.848-0.862) when diagnosing without AI to 0.885 (95% CI 0.879-0.891; P < .0001) for the senior radiologists, and from 0.819 (95% CI 0.811-0.826) to 0.866 (95%CI 0.859-0.872; P < .0001) for the junior radiologist. In the simulated scenario of clinical reading using images and videos in test C, the pooled AUROC of the final diagnosis with AI assistance was improved to 0.873 (95% CI 0.863-0.83) from the 0.862 (95% CI 0.851-0.872; p < .0001) of radiologists reviewing videos and images without AI assistance. The details of training and validation for AI model establishment can be found in previous article¹. The AI model is available via the web page http://119.91.96.49:8686/ (established by Maiying Technology Co., Ltd, Guangzhou, China).

eReference.

 Peng S, Liu Y, Lv W, et al. Deep learning-based artificial intelligence model to assist thyroid nodule diagnosis and management: a multicentre diagnostic study. *The Lancet Digital health*. 2021;3(4):e250-e259. doi:10.1016/s2589-7500(21)00041-8

		Retrospective set (n=1754)	Prospective set (n=300)
Sex	Male	439 (25.0)	82 (27.3)
	Female	1315 (75.0)	218 (72.7)
Age (years)	Mean (SD)	42.1 (13.2)	41.7 (14.1)
	Median (IQR)	41 (33-52)	40 (32-53)
			54 (40.0)
	TR 1	95 (5.4)	54 (18.0)
	TR 2	112 (6.4)	48 (16.0)
ACR TI-RADS category	TR 3	229 (13.1)	6 (2.0)
	TR 4	459 (26.2)	34 (11.3)
	TR 5	859 (49.0)	158 (52.7)

eTable 2. The Demographic Characteristics of Patients and the Nodules Distribution of ACR TI-RADS Category

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data System; IQR, inter-quartile range; SD, standard deviation; TR, TI-RADS category.

Sex and ACR TI-RADS category are listed as numbers of images, while age is listed as the number of patients.

Except where indicated, data in parentheses are percentages.

	Retrospectiv (n=1754)	e set	Prospective (n=300)	set
Variable	Benign (n=748)	Malignant (n=1006)	Benign (n=125)	Malignant (n=175)
ACR TI-RADS features				
Composition				
Cystic or almost completely cystic	60 (8.0)	0 (0.0)	30 (24.0)	0 (0.0)
Spongiform	35 (4.7)	0 (0.0)	24 (19.2)	0 (0.0)
Mixed cystic and solid	167 (22.6)	6 (0.6)	57 (45.6)	5 (2.9)
Solid or almost completely solid	486 (65.0)	1000 (94.0)	14 (11.2)	170 (97.1)
Echogenicity				
Anechoic	60 (8.0)	0 (0.0)	30 (24.0)	0 (0.0)
Hyperechoic or isoechoic	354 (47.3)	29 (2.7)	65 (52.0)	6 (3.4)
Hypoechoic	327 (43.7)	792 (74.4)	24 (19.2)	117 (66.9)
Very hypoechoic	7 (0.9)	185 (17.4)	6 (4.8)	52 (29.7)
Shape				
Wider-than-tall	712 (95.2)	594 (55.8)	123 (98.4)	99 (56.6)
Taller-than-wide	36 (4.8)	412 (38.7)	2 (1.6)	76 (43.4)
Margin				
Smooth	333 (44.5)	45 (4.2)	94 (75.2)	8 (4.6)
III-defined	379 (50.7)	587 (55.2)	28 (22.4)	76 (43.4)
Lobulated or irregular	35 (4.7)	340 (32.0)	3 (2.4)	73 (41.7)
Extra-thyroidal extension	1 (0.1)	34 (3.2)	0 (0.0)	18 (10.3)
Echogenic foci				
None or large comet-tail artifacts	619 (82.8)	394 (37.0)	110 (88.0)	46 (26.3)
Macrocalcifications	62 (8.3)	144 (13.5)	9 (7.2)	42 (24.0)
Peripheral (rim) calcifications	13 (1.7)	13 (1.2)	2 (1.6)	3 (1.7)
Punctate echogenic foci	70 (9.4)	522 (49.1)	9 (7.2)	107 (61.1)
Nodule size				
< 5 mm	18 (2.4)	63 (5.9)	6 (4.8)	13 (7.4)
5 mm-10 mm	94 (12.6)	400 (37.6)	11 (8.8)	75 (42.9)
10 mm-20 mm	242 (32.4)	427 (40.1)	30 (24.0)	61 (34.9)
> 20 mm	394 (52.7)	116 (10.9)	78 (62.4)	26 (14.9)
Parenchyma				
Normal parenchyma	507 (67.8)	739 (69.5)	82 (65.6)	103 (58.9)
Diffused parenchyma	241 (32.2)	267 (25.1)	43 (34.4)	72 (41.1)

eTable 3. Distribution of Images Features

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data

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System. Data are number of nodules, with percentages in parentheses.

eTable 4. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Composition Features in Junior and Senior Radiologists

	Sensitivity			Specificity		
	without	with	- p	without	with	- p value
	AI (%)	AI (%)	value	AI (%)	AI (%)	
Cystic or almost completely cystic						
Junior				98 (96, 100)	97 (95, 99)	.55
Senior				99 (97, 100)	99 (97, 100)	>.99
Spongiform						
Junior				91 (87, 95)	93 (89, 97)	.39
Senior				93 (89, 97)	94 (91, 98)	.73
Mixed cystic and						
solid						
Junior	37 (18, 55)	40 (21, 59)	>.99	92 (91, 94)	96 (95, 97)	<.001
Senior	57 (38, 75)	63 (45, 82)	.63	95 (94, 97)	97 (96, 98)	.001
Solid or almost						
completely solid						
Junior	89 (88, 89)	93 (92, 93)	<.001	67 (65, 69)	73 (71, 75)	<.001
Senior	91 (90, 91)	93 (92, 94)	<.001	72 (70, 74)	75 (74, 77)	<.001

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data

System; AI, Artificial intelligence.

	Sensitivity			Specificity			
	without	with	– p value	without	with	- p value	
	AI (%)	AI (%)	value	AI (%)	AI (%)	value	
Anechoic							
Junior				98 (96, 100)	97 (95, 99)	.55	
Senior				99 (97, 100)	99 (97, 100)	>.99	
Hyperechoic							
or isoechoic							
Junior	44 (36, 52)	55 (47, 63)	.001	90 (89, 92)	94 (93, 95)	<.001	
Senior	43 (35, 52)	57 (49, 65)	.001	94 (93, 95)	96 (95, 97)	.002	
Hypoechoic							
Junior	89 (88, 90)	93 (92, 94)	<.001	59 (56, 61)	65 (63, 67)	<.001	
Senior	91 (90, 92)	93 (93, 94)	<.001	64 (61, 66)	68 (65, 70)	<.001	
Very							
hypoechoic							
Junior	93 (92, 95)	97 (96, 98)	<.001	17 (4, 30)	23 (8, 37)	.75	
Senior	96 (95, 97)	96 (95, 98)	.68	14 (2, 26)	34 (18, 51)	.07	

eTable 5. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Echogenicity Features in Junior and Senior Radiologists

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data System; AI, Artificial intelligence.

eTable 6. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Shape Features in Junior and Senior Radiologists

	Sensitivity			Specificity		
	without	with	- p value	without	with	– p value
	AI (%)	AI (%)	. and o	AI (%)	AI (%)	laide
Wider-than-tall						
Junior	84 (82, 85)	89 (88, 90)	<.001	79 (78, 80)	84 (82, 85)	<.001
Senior	86 (85, 87)	90 (89, 90)	<.001	83 (82, 84)	86 (84, 87)	<.001
Taller-than-wide	e					
Junior	95 (94, 96)	97 (96, 98)	<.001	24 (18, 31)	28 (21, 34)	.29
Senior	96 (95, 97)	97 (96, 98)	.08	31 (24, 37)	33 (26, 40)	.36

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data System; AI, Artificial intelligence.

	Sensitivity	Sensitivity		Specificity		
	without	with	— р	without	with	– p value
	AI (%)	AI (%)	value	AI (%)	AI (%)	
Smooth						
Junior	45 (38, 51)	60 (53, 66)	<.001	93 (91, 94)	94 (93, 95)	.08
Senior	46 (40, 53)	57 (51, 64)	<.001	95 (94, 96)	96 (95, 97)	.033
III-defined						
Junior	87 (86, 88)	92 (91, 93)	<.001	68 (66, 70)	75 (73, 77)	<.001
Senior	90 (89, 91)	93 (92, 94)	<.001	74 (72, 76)	78 (76, 79)	<.001
Lobulated or						
irregular						
Junior	95 (94, 96)	97 (96, 97)	.001	14 (9, 20)	18 (13, 24)	.14
Senior	97 (96, 97)	97 (96, 98)	.65	19 (13, 25)	22 (16, 28)	.41
Extra-thyroid						
al extension						
Junior	95 (91, 98)	99 (97, 100)	.004	0	60 (-8, 128)	.25
Senior	98 (95, 100)	98 (96, 100)	>.99	0	40 (-3, 108)	.50

eTable 7. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Margin Features in Junior and Senior Radiologists

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data

System; AI, Artificial intelligence.

	Sensitivity			Specificity		
	without	with	- p	without	with	- p
	AI (%)	AI (%)	value	AI (%)	AI (%)	value
None or large						
comet-tail artifacts						
Junior	80 (78, 82)	87 (86, 89)	<.001	84 (83, 86)	87 (86, 88)	<.001
Senior	85 (83, 87)	87 (86, 89)	<.001	87 (86, 89)	90 (88, 91)	<.001
Macrocalcifications						
Junior	91 (89, 93)	95 (93, 96)	<.001	42 (36, 47)	52 (46, 57)	<.001
Senior	91 (89, 93)	95 (94, 97)	<.001	54 (48, 59)	56 (51, 62)	.32
Peripheral (rim)						
calcifications						
Junior	69 (58, 81)	77 (66, 87)	.13	57 (45, 69)	60 (48, 72)	.77
Senior	60 (48, 72)	79 (68, 89)	.004	68 (56, 79)	60 (48, 72)	.23
Punctate echogenic						
foci						
Junior	95 (94, 95)	97 (96, 97)	<.001	33 (28, 38)	46 (40, 51)	<.001
Senior	95 (94, 96)	97 (96, 98)	<.001	38 (33, 43)	45 (40, 50)	<.001
20				50 (00, 10)		

eTable 8. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different ACR TI-RADS Echogenic Foci Features in Junior and Senior Radiologists

Abbreviations: ACR TI-RADS, The American College of Radiology published the Thyroid Imaging Reporting and Data System; AI, Artificial intelligence.

	Sensitivity			Specificity		
	without	with	- p value	without	with	– p value
	AI (%)	AI (%)		AI (%)	AI (%)	
< 5 mm						
Junior	86 (82, 90)	89 (85, 92)	.14	68 (58, 78)	67 (57, 77)	>.99
Senior	89 (86, 93)	92 (89, 95)	.11	64 (54, 75)	73 (64, 83)	.039
5 mm-10 mm						
Junior	88 (87, 90)	93 (92, 94)	<.001	57 (53, 62)	64 (59, 68)	<.001
Senior	91 (90, 93)	93 (92, 94)	.001	64 (60, 68)	66 (61, 70)	.33
10 mm-20 mm						
Junior	90 (89, 91)	95 (94, 95)	<.001	68 (65, 71)	74 (71, 76)	<.001
Senior	92 (91, 93)	95 (94, 96)	<.001	73 (70, 75)	76 (74, 78)	<.001
> 20 mm						
Junior	82 (79, 85)	86 (83, 89)	<.001	86 (85, 88)	90 (89, 91)	<.001
Senior	81 (78, 84)	84 (81, 87)	.009	90 (89, 91)	92 (91. 93)	<.001

eTable 9. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different Nodule Sizes in Junior and Senior Radiologists

Abbreviations: AI, Artificial intelligence.

eTable 10. Comparison of Independent Diagnosis and AI Model Assistance in Sensitivity and Specificity Under Different Parenchymal Backgrounds in Junior and Senior Radiologists

	Sensitivity			Specificity	Specificity	
	without	with	- p	without	with	- p
	AI (%)	AI (%)	value	AI (%)	AI (%)	value
Homogeneous						
parenchyma						
Junior	88 (87, 89)	92 (91, 93)	<.001	76 (74, 77)	80 (78, 81)	<.001
Senior	90 (89, 91)	93 (92. 93)	<.001	79 (78, 81)	82 (81, 84)	<.001
Heterogeneous						
parenchyma						
Junior	90 (88, 91)	94 (93, 95)	<.001	77 (75, 80)	83 (81, 85)	<.001
Senior	90 (89, 92)	94 (93, 95)	<.001	83 (81, 85)	85 (83, 87)	.028

Abbreviations: AI, Artificial intelligence.



eFigure 1. The Significant Features for Junior Radiologists

For junior radiologists, mixed cystic and solid, solid or almost completely solid, hyperechoic or isoechoic, hypoechoic, very hypoechoic, wider-than-tall, taller-than-wide, smooth, ill-defined, lobulated or irregular, extra-thyroidal extension, none or large comet-tail artifacts, macrocalcifications, punctate echogenic foci, homogeneous and heterogeneous parenchyma, and nodule sizes \geq 5mm were investigated as significant features.



eFigure 2. The Significant Features for Senior Radiologists

For senior radiologists, mixed cystic and solid, solid or almost completely solid, hyperechoic or isoechoic, hypoechoic, wider-than-tall, smooth, ill-defined, none or large comet-tail artifacts, macrocalcifications, punctate echogenic foci, homogeneous and heterogeneous parenchyma, and all nodule sizes were investigated as significant features.