

# Test Report

## VUNO Med®-Chest X-ray™

โดย บริษัท ไทย จีแอล จำกัด

รายงานผลการทดสอบ โดยราชวิทยาลัยรังสีแพทย์แห่งประเทศไทย

ทดสอบใช้กับภาพรังสีทรวงอก ในกรณี

- คัดกรอง (screening) วัณโรคปอด
  อ่านผลซ้ำ (double reading) ให้กับรังสีแพทย์ เพื่อเพิ่มคุณภาพการวินิจฉัย
  เพิ่มความแม่นยำในการค้นหาพยาธิสภาพให้กับรังสีแพทย์
- ประมาณความยาก-ง่ายในการแปลผล



## Report on the Test Performance of Artificial Intelligence for Tuberculosis Screening in Chest X-Ray Images of the Thai Population

## Filer Name

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## **Developer Company**

Company	Vuno, Inc	
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Country	South Korea	
Website	https://www.vuno.co	

## Software

Name	VUNO Med®-Chest X-ray™
Version	V1.3
Descriptio	Product specifications excerpted from <a href="https://grand-challenge.org/aiforradiology/">https://grand-challenge.org/aiforradiology/</a>
n	WINO Med® Chest X-Ray™ is a deep learning-based screening solution      for five major lung diseases - Nodule/Mass, Consolidation, Interstitial Opacity,      Pneumothorax, Pleural Effusion - on chest X-ray (PA/AP) images. This algorithm      provides information on the presence of the abnormalities, their names, abnormality      scores, and locations.



## The Development of the Dataset for Testing Artificial Intelligence for Tuberculosis Screening in Chest X-Ray Images of the Thai Population

Product specifications	Information source: Vendor Last updated: Oct. 28, 2023
	General
Product name	VUNO Med®-Chest X-ray™
Company	VUNO
Subspeciality	Chest
Modality	X-ray
Disease targeted	Nodule/Mass, Consolidation, Interstitial Opacity, Pneumothorax, Pleural Effusion
Key-features	Abnormality detection
	Data characteristics
Population	All population with a risk of thoracic abnormalities
Input	Chest XR PA/AP images
Input format	DICOM
Output	Abnormality score, Lesion heatmap, Lesion boundary
Output format	DICOM, GSPS
	Technology
Integration	Integration in standard reading environment (PACS), Integration via AI marketplace or distribution platform, Stand-alone third party application, Stand-alone webbased
Deployment	Locally on dedicated hardware, Locally virtualized (virtual machine, docker), Cloud-based, Hybrid solution
Trigger for analysis	Automatically, right after the image acquisition
Processing time	< 3 sec
	Certification
CE	Certified, Class IIa, MDD
FDA	No or not yet
	Market presence
On market since	06-2020
Distribution channels	Tempus Pixel, Samsung Electronics
Countries present	10+
	Pricing
Pricing model	Pay-per-use, Subscription
Based on	Number of analyses



#### Dataset

Reference No.	1A2A
Number of Images	808
Internal Validation	Consistent

## **Data Characteristics**

The dataset consists of 808 randomly selected chest radiographic images from a pool of 1,500 images carefully curated from Songklanagarind Hospital in Songkhla Province, Chiangrai Pracharuk Hospital in Chiang Rai Province, Udon Thani Hospital in Udon Thani Province, Suttawet Hospital in Maha Sarakham Province, and the Tuberculosis Division of the Department of Disease Control, Ministry of Public Health. Each image was read by three B Readers. Our goal is to utilize high-quality datasets that are read by B Readers, who are trained and certified radiologists.

A B Reader is a qualified radiologist who is certified by the National Institute for Occupational Safety and Health (NIOSH) in the United States. B Readers are specifically trained to interpret and classify chest radiographs for the presence of pneumoconiosis, a group of lung diseases.

Characteristics of the radiographic images:

- Chest radiographic images of patients aged 15 years and above were included, taken with a computed radiography machine.
- No images from patients with a positive HIV Serology status.
- No images from patients with other opportunistic pulmonary infections or co-infections, such as Mycobacterium tuberculosis, Histoplasmosis, Cryptococcosis, Melioidosis, and Acinetobacter baumannii.

To assess the inter-rater reliability, the following metrics were employed:

- Pairwise Agreement: The average level of agreement among each pair of B readers.
- Intraclass Agreement (ICC): The average Pearson's correlation using ICC(2,3) when three B readers read the randomly selected radiographic images.
- Pairwise Cohen's Kappa and Fless' Kappa statistics for the analysis of agreement between assessors

## **Number of Findings**

Table 1 presents the number of findings annotated by B Readers for chest X-ray images in Dataset 1A2A, which consists of 808 images. Each image in the dataset was independently assessed by three randomly selected B Readers from a pool of six B Readers.  $N_{Individual Reader}$  represents The number of findings that each individual B reader labelled, while  $N_{Concensus}$  represents the number of findings where the majority of the B Readers agreed.



Finding	<b>N</b> Individual Reader	N <sub>Concensus</sub>
Abnormalities	1,575	513
Small opacity	1,252	421
Primary nodular	929	324
Primary reticular	308	58
Secondary nodular	718	242
Secondary reticular	455	110
Large opacity	1,240	422
Mass/nodule	497	136
Cavity	881	298
Fibrosis	742	243
Calcification	299	58
Pleural effusion	327	109
Pleural thickening	556	179
Pneumothorax	14	4
Hilar adenopathy	316	72
Mediastinal adenopathy	96	17
Consistent with tuberculosis	1,270	416
Active Tuberculosis	1,222	408
Patchy infiltration	930	336
Cavity with surrounding		
consolidation	813	280
Unilateral hilar/paratracheal		
lymph node enlargement	147	30
Pleural effusion	165	49
Miliary nodules	310	76
Indeterminate tuberculosis	48	6
Reticulonodular infiltration	28	4
Destroyed lung or		
bronchiectasis	5	0
Inconsistent with tuberculosis	1,154	392

Table 1 Number of findings annotated annotated by B Readers in Dataset 1A2A





## Inter-rater Reliability

<u>Table 2</u> Inter-rater reliability measures for each finding in Dataset 1A2A (808 images). Each finding was interpreted by three B Readers. The reliability was measured using statistical metrics such as Pairwise Agreement, ICC(2,3), Pairwise Cohen's kappa, and Fleiss' kappa.

Finding		Finding	Agreement	ICC	Cohen's	Fleiss'
А	Abnormalities		0.9208	0.9345	0.826	0.826
S	ma	ll opacity	0.8589	0.8841	0.7175	0.7175
	P	rimary nodular	0.8276	0.8395	0.6352	0.6352
	P	rimary reticular	0.8069	0.3092	0.1296	0.1297
	S	econdary nodular	0.7063	0.5576	0.2953	0.2955
	S	econdary reticular	0.7434	0.3615	0.1587	0.1585
L	arg	e opacity	0.9043	0.9269	0.8085	0.8085
N	lass	s/nodule	0.7748	0.5734	0.309	0.309
С	avi	ty	0.8688	0.8837	0.7168	0.7165
F	ibrc	osis	0.7632	0.7051	0.4429	0.4426
С	alci	ification	0.8177	0.3586	0.157	0.1569
Ρ	leu	ral effusion	0.9389	0.8945	0.7381	0.7384
Ρ	leu	ral thickening	0.8457	0.7951	0.5639	0.5636
Ρ	neı	ımothorax	0.9967	0.8816	0.7095	0.7126
Н	ilar	adenopathy	0.8399	0.5564	0.2952	0.294
N	ledi	astinal adenopathy	0.9398	0.4438	0.2062	0.2082
С	ons	sistent with tuberculosis	0.9604	0.9721	0.9206	0.9206
	A	ctive Tuberculosis	0.9538	0.9672	0.9076	0.9076
		Patchy infiltration	0.8284	0.8407	0.6371	0.6371
		Cavity with surrounding consolidation	0.8507	0.8565	0.6651	0.665
		Unilateral hilar/paratracheal lymph node enlargement	0.9051	0.3763	0.1651	0.1672
		Pleural effusion	0.9406	0.7733	0.5314	0.5318
		Miliary nodules	0.8234	0.4418	0.2087	0.2084
Indeterminate tuberculosis		determinate tuberculosis	0.9686	0.4183	0.197	0.1923
		Reticulonodular infiltration	0.9802	0.3178	0.1643	0.1328
		Destroyed lung or bronchiectasis	0.9959	-0.006	-0.0019	-0.0021
lr	icor	nsistent with tuberculosis	0.9604	0.9721	0.9206	0.9206



ICC/Kappa Statistic	Strength of Agreement
<0.00	Poor
0.00 – 0.20	Slight
0.21 – 0.40	Fair
0.41 - 0.60	Moderate
0.61 – 0.80	Substantial
0.81 – 1.00	Almost Perfect

Table 3 Interpretation of ICC and Kappa Values according to Landis and Koch (1977)<sup>1</sup>

<sup>1</sup> Landis, J. R., & Koch, G. G. (1977). The Measurement of Observer Agreement for Categorical Data. In Biometrics (Vol. 33, Issue 1, p. 159). JSTOR. https://doi.org/10.2307/2529310

## Results

The inter-rater reliability is measured using Pairwise Agreement, which is the average similarity between each pair of B Readers and VUNO Med-CXR, as well as Pairwise Cohen's Kappa, which is the average of Cohen's Kappa statistics between each pair of B Readers and VUNO Med-CXR. This is done to compare the agreement between B Readers and VUNO Med-CXR ("B" vs Al) and among B Readers themselves ("B" vs "B").

<u>Table 4</u> Reliability Measures Within B Readers ("B" vs "B") and Between the System and B Readers ("B" vs AI)

Finding	N	Threshold	Pairwise Agreement		Pairwise Agreement Cohen's Kap		Карра
			"B" vs "B"	"B" vs Al	"B" vs "B"	"B" vs Al	
Abnormality	1,575	0.65	0.9225	0.9187	0.8296	0.8222	
Tuberculosis(Active)	1,222	0.25	0.9538	0.8932	0.9060	0.7865	

For measuring the diagnostic performance of each disease annotation, criteria such as Sensitivity, Specificity, Positive Prediction Rate (PPR), and Negative Prediction Rate (NPR) are utilized. These metrics are evaluated using the diagnostic threshold specified by the manufacturer, along with the area under the ROC curve.

Table 5 Diagnostic Performance of Each Finding by the System Compared to B Readers

Finding	N	Threshold	Sensitivity	Specificity	PPV	NPV	AUROC
Abnormality	1,575	0.65	0.9327	0.8928	0.9417	0.8773	0.9660
Tuberculosis(Active)	1,222	0.25	0.8249	0.9626	0.9573	0.8439	0.9655



Figure 1 ROC Curves Illustrating Diagnostic Performance for Each Finding



## **Analysis of Results**

According to Table 6, when comparing Pairwise Agreement and Cohen's Kappa between B Readers and VUNO Med-CXR ("B" vs AI) and among B Readers themselves ("B" vs "B"), VUNO Med-CXR demonstrates performance close to that of B Readers (with a difference of less than 5%). For abnormalities, the Pairwise Agreement of among B readers scored higher than the Pairwise Agreement of each B reader and VUNO Med-CXR by 0.38% (N=1,575) and the Cohen's Kappa of among B readers scored higher than the Cohen's Kappa of each B reader and VUNO Med-CXR by 0.74% (N=1,575). For tuberculosis, the agreement of among B readers scored higher than the agreement of each B reader and VUNO Med-CXR by 7.69% (N=1,222).

Table 6 Differences between Pairwise Agreement and Cohen's Kappa

Finding	Pairwise Agreement			Cohen's Kappa		
	B vs "B"	"B" vs Al	Diff	"B" vs "B"	"B" vs Al	Diff
Abnormality	0.9225	0.9187	-0.38%	0.8296	0.8222	-0.74%
Tuberculosis(Active)	0.9602	0.8833	-7.69%	0.9186	0.7865	-13.21%



Regarding the lung tuberculosis screening, VUNO Med-CXR, when analyzed on Dataset 1A2A showed diagnostic performance closely comparable to that of B Readers. It achieved an area under the receiver operating characteristic curve (AUROC) of 0.9655, sensitivity of 0.8249, and specificity of 0.9626 at a threshold of 0.25.

Referring to <u>The Target Product Profiles (TPPs) for a rapid non-sputum-based biomarker test</u> <u>for tuberculosis detection</u> by the World Health Organization (WHO), as shown in Table 7, it can be observed that each test scenario has different criterions for sensitivity and specificity.

Table 7 TPP for a rapid non-sputum-based biomarker test for tuberculosis detection

	Minimal Requi	rements	Optimal Requirements		
	Sensitivity	Specificity	Sensitivity	Specificity	
Smear-replacement test	Overall >80%	98%	Overall >95%	98%	
	Positive >99%		Positive >99%		
	Negative >60%		Negative >68%		
Non-sputum based	Overall >65%	98%	Positive >98%	98%	
biomarker test	Positive >98%		Negative >68%		
Triage test	90%	70%	95%	80%	

Reference: https://academic.oup.com/jid/article/211/suppl\_2/S29/2490781

The Minimal Requirements and Optimal Requirements in the WHO TPPs (Target Product Profiles) outline the minimum and ideal thresholds for sensitivity and specificity that such a test should meet.

The Minimal Requirements indicate the minimum acceptable level of sensitivity and specificity that the test should achieve to be considered effective for tuberculosis detection. These criteria serve as a baseline standard for performance.

The Optimal Requirements represent the desired ideal performance levels for sensitivity and specificity. Meeting or exceeding these requirements would indicate a highly accurate and reliable test for tuberculosis detection.

The results of tuberculosis screening using VUNO Med-CXR at different thresholds compared to the WHO TPP criteria, with the highest threshold that yields the closest specificity to the WHO TPP, are presented in Table 8.



Threshold	Sensitivity	Specificity
0.3272	0.7758	0.9800
0.2512	0.8249	0.9626
0.0300	0.9640	0.8020
0.0101	0.9812	0.6448

Table 8 Sensitivity and Specificity Values at Different Thresholds according to WHO TPP Criteria

Furthermore, when comparing the results obtained with the WHO TPP criteria, it was found that VUNO Med-CXR met the requirements for the Triage test (for both the Minimal Requirements and Optimal Requirements) and the Non-sputum based biomarker test (for the Minimal Requirements criteria). The test outcomes are summarized in Table 9.

Table 9 Results of Tuberculosis Screening by VUNO Med-CXR according to WHO TPP Criteria.

	Minimal Requirements	<b>Optimal Requirements</b>
Smear-replacement test	Not Pass	Not pass
Non-sputum based biomarker test	Pass	Not pass
Triage test	Pass	Pass



## Supplementary Table

Figure 1 illustrates the Receiver Operating Characteristic (ROC) curve which can be used to visualize the performance of a classifier at various thresholds. By adjusting the threshold, one change the trade-off between sensitivity and specificity. Table S1 and S2 details different sensitivity and specificity values across varying classification thresholds for abnormalities and tuberculosis, respectively.

Threshold	Sensitivity	Specificity
0.9900	0.5238	0.9965
0.9773	0.6825	0.9941
0.9505	0.8178	0.9835
0.9271	0.8483	0.9764
0.9080	0.8711	0.9694
0.8766	0.8965	0.9635
0.8532	0.9016	0.9517
0.8321	0.9111	0.9517
0.8046	0.9143	0.9435
0.7761	0.9213	0.9317
0.7613	0.9232	0.9282
0.7266	0.9257	0.9081
0.7034	0.9263	0.9022
0.6765	0.9276	0.8975
0.6504	0.9327	0.8928
0.6254	0.9384	0.8893
0.6149	0.9397	0.8881
0.5776	0.9410	0.8728
0.5525	0.9435	0.8598
0.5274	0.9467	0.8516

<u>Table S1</u> Sensitivity and Specificity Across Varying Classification Thresholds for Abnormalities. (Manufacturer's recommended threshold value is 0.65)

	1	1
Threshold	Sensitivity	Specificity
0.5044	0.9511	0.8386
0.4844	0.9524	0.8269
0.4698	0.9530	0.8245
0.4202	0.9530	0.8033
0.4000	0.9543	0.7915
0.3832	0.9549	0.7856
0.3515	0.9575	0.7338
0.3289	0.9619	0.7244
0.3059	0.9632	0.7020
0.2757	0.9632	0.6914
0.2513	0.9663	0.6584
0.2259	0.9689	0.6243
0.2107	0.9708	0.6066
0.1755	0.9733	0.5548
0.1560	0.9778	0.5383
0.1255	0.9790	0.4876
0.1082	0.9816	0.4429
0.0759	0.9873	0.3439
0.0524	0.9917	0.2674
0.0259	0.9975	0.0978



Threshold	Sensitivity	Specificity
0.9900	0.1465	0.9992
0.9601	0.3126	0.9983
0.8525	0.4722	0.9983
0.8099	0.5082	0.9925
0.7581	0.5426	0.9925
0.7027	0.5761	0.9917
0.6833	0.5827	0.9908
0.6799	0.5827	0.9884
0.6276	0.6268	0.9884
0.6143	0.6268	0.9859
0.5512	0.6538	0.9859
0.5018	0.6931	0.9834
0.4383	0.7267	0.9825
0.3272	0.7758	0.9800
0.3050	0.7856	0.9775
0.2876	0.8028	0.9750
0.2512	0.8249	0.9626
0.2339	0.8339	0.9592
0.9900	0.1465	0.9992
0.9601	0.3126	0.9983

<u>Table S2</u> Sensitivity and Specificity Across Varying Classification Thresholds for Tuberculosis. (Manufacturer's recommended threshold value is 0.25)

Threshold	Sensitivity	Specificity
0.2086	0.8478	0.9559
0.1878	0.8576	0.9534
0.1662	0.8576	0.9509
0.1588	0.8650	0.9509
0.1305	0.8748	0.9384
0.1179	0.8871	0.9334
0.1100	0.8920	0.9285
0.1007	0.9051	0.9243
0.0915	0.9108	0.9201
0.0836	0.9108	0.9126
0.0704	0.9198	0.8943
0.0612	0.9345	0.8844
0.0533	0.9452	0.8627
0.0409	0.9484	0.8411
0.0300	0.9640	0.8020
0.0202	0.9714	0.7521
0.0101	0.9812	0.6448
0.0052	0.9877	0.5216
0.2086	0.8478	0.9559
0.1878	0.8576	0.9534

Typically, at a lower threshold, the model has high sensitivity but lower specificity. This means it correctly identifies most of the positives but also produces more false positives. At a medium threshold, there's a balance between sensitivity and specificity, which might be a good choice depending on the context. At a higher threshold, the model has high specificity but lower sensitivity. This is suitable when you want to be very certain about the positives but risk missing some.

Choosing the optimal threshold depends on the specific requirements of the task at hand. In some applications, high sensitivity might be more important, while in others, high specificity may be preferred.

24<sup>th</sup> February 2024