

LUNG CANCER SCREENING AND INCIDENTALS

LUNG CANCER

Taking it to the next level:

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IS THERE A ROLE FOR ARTIFICIAL INTELLIGENCE?



DISCLOSURES

MEDTRONIC ROBOTICS ADVISORY BOARD
ASTRA ZENECA SPEAKERS BUREAU

i3 Health and FLASCO have mitigated all relevant financial relationships

EARLY

What are we trying to do?

DETECTION

Assess Risk to Detect Early Cancers

ARTIFICIAL INTELLIGENCE



CONVERT DATA  INTO INSIGHTS  INTO ACTION

EARLY
**Can Artificial Intelligence disrupt current
methodology?**
DETECTION

YES, WITH CAVEATS....

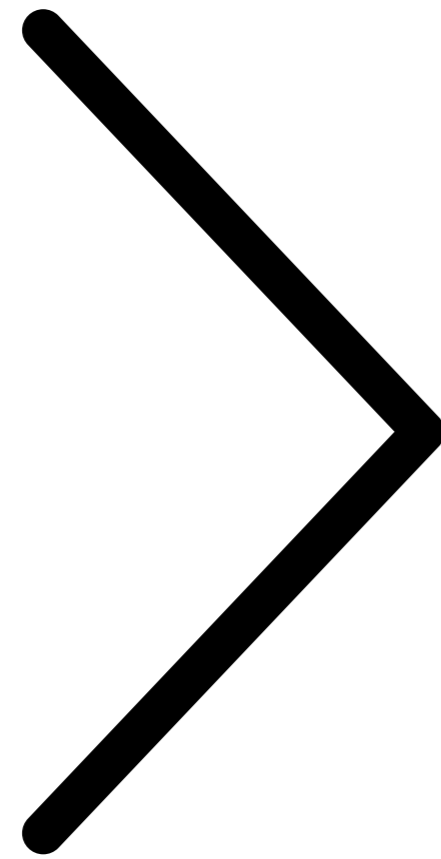
AI HELPS
ASSESS
RISK
FOR
LUNG
CANCER

IDENTIFY RISK

STRATIFY RISK

ANALYZE RISK

MITIGATE RISK

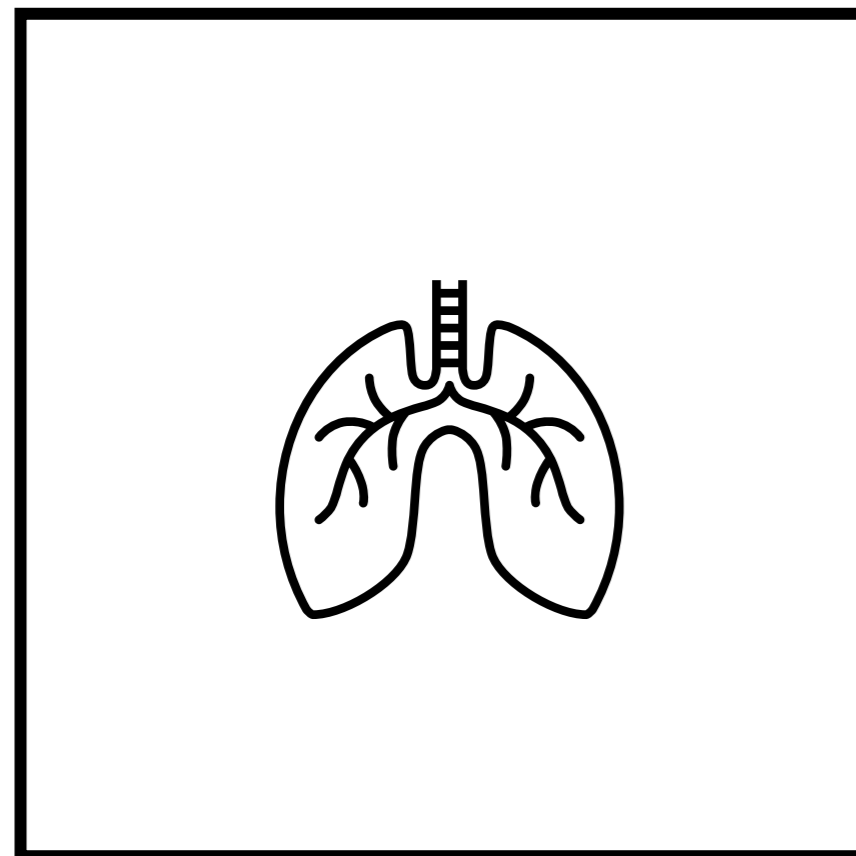


EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

IDENTIFY RISK

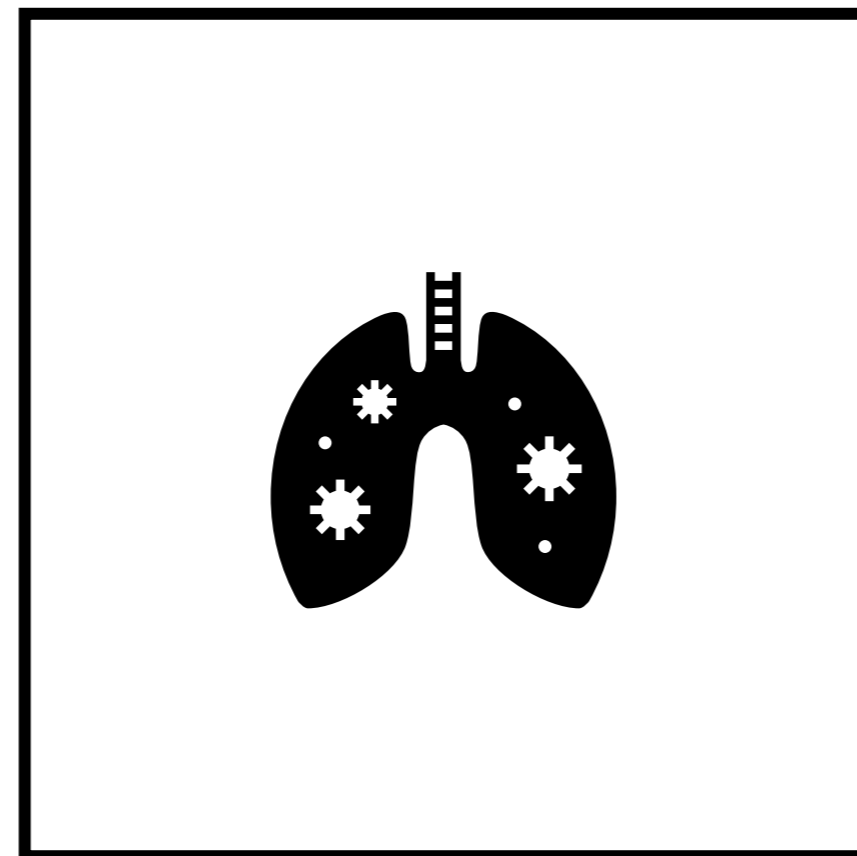
DATA MINING AND NODULE IDENTIFICATION

CLINICAL RISK PREDICTORS ARE LIMITED AND LABORIOUS



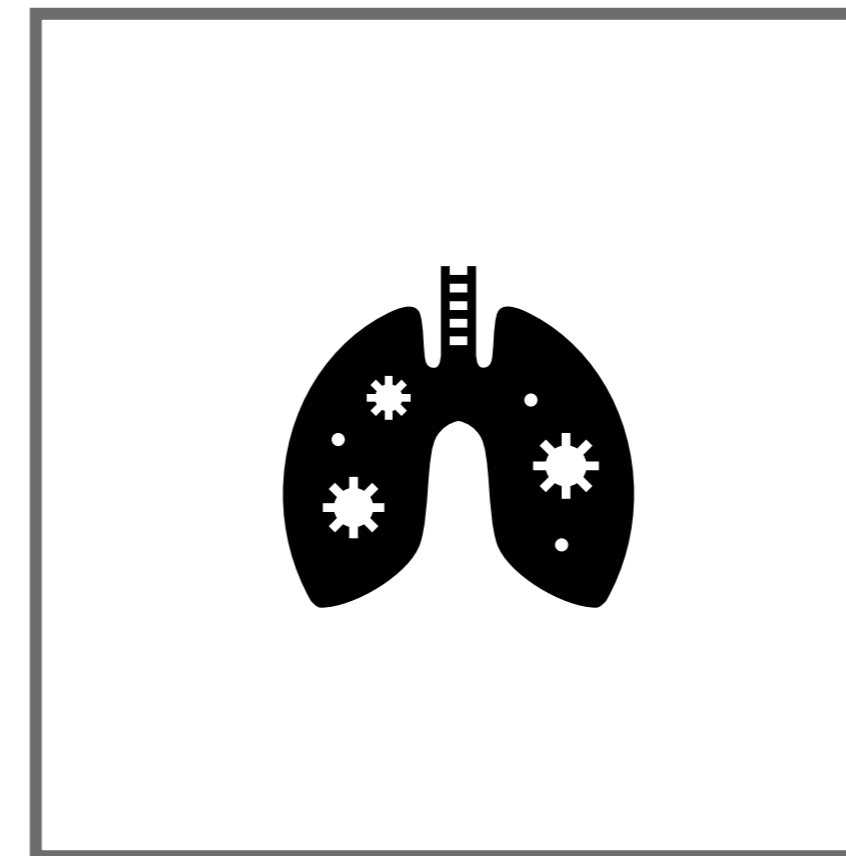
PLCO M2012

SOCIODEMOGRAPHIC,
MEDICAL HISTORY,
FOUR SMOKING
VARIABLES, AGE,
FAMILY HISTORY, ETC.



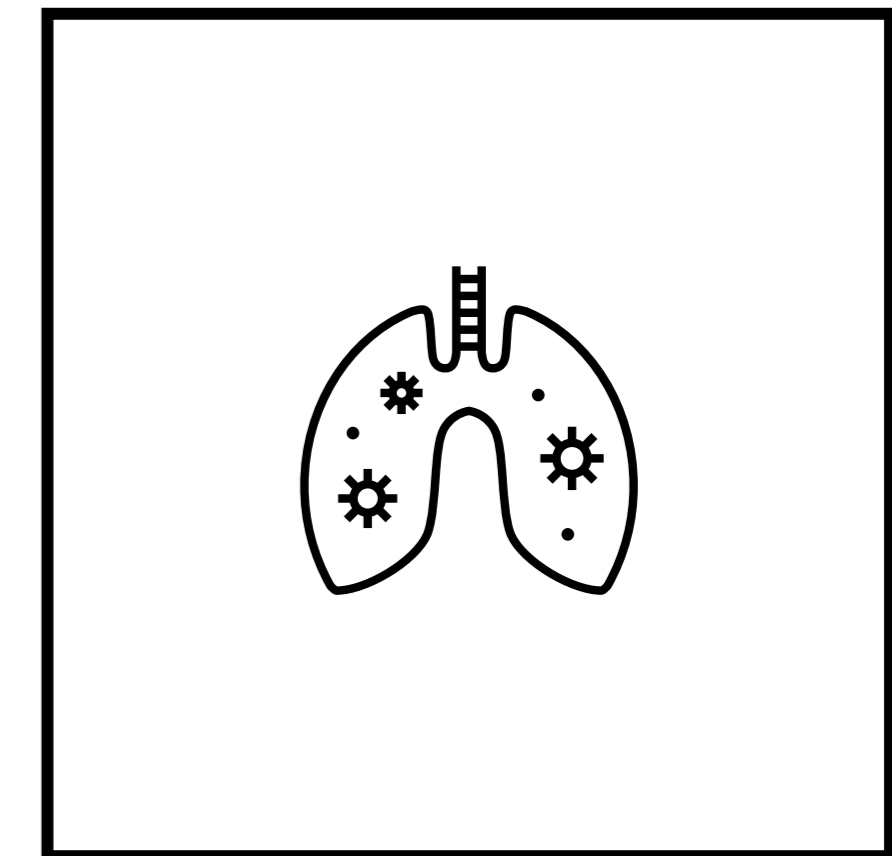
BROCK, MAYO

REQUIRES SIGNIFICANT
CLINICAL DATA



LUNG RADS

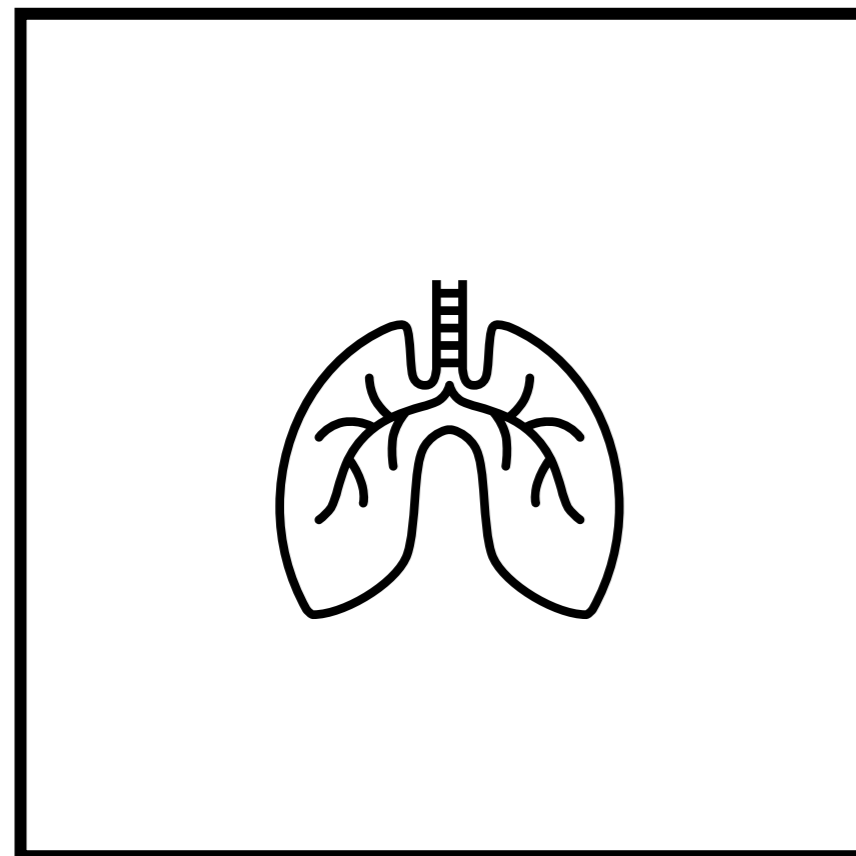
REQUIRES NODULE:
CHARACTERISTICS, SIZE,
CHANGE



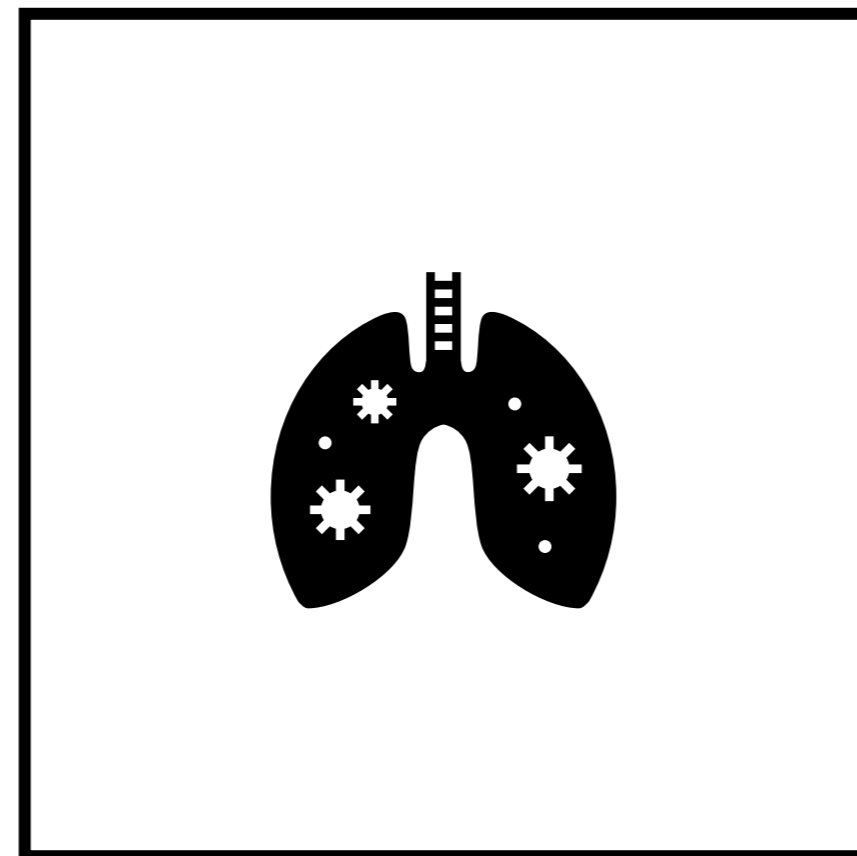
FLESCHNER'S

REQUIRES NODULE:
CHARACTERISTICS, SIZE,
CHANGE

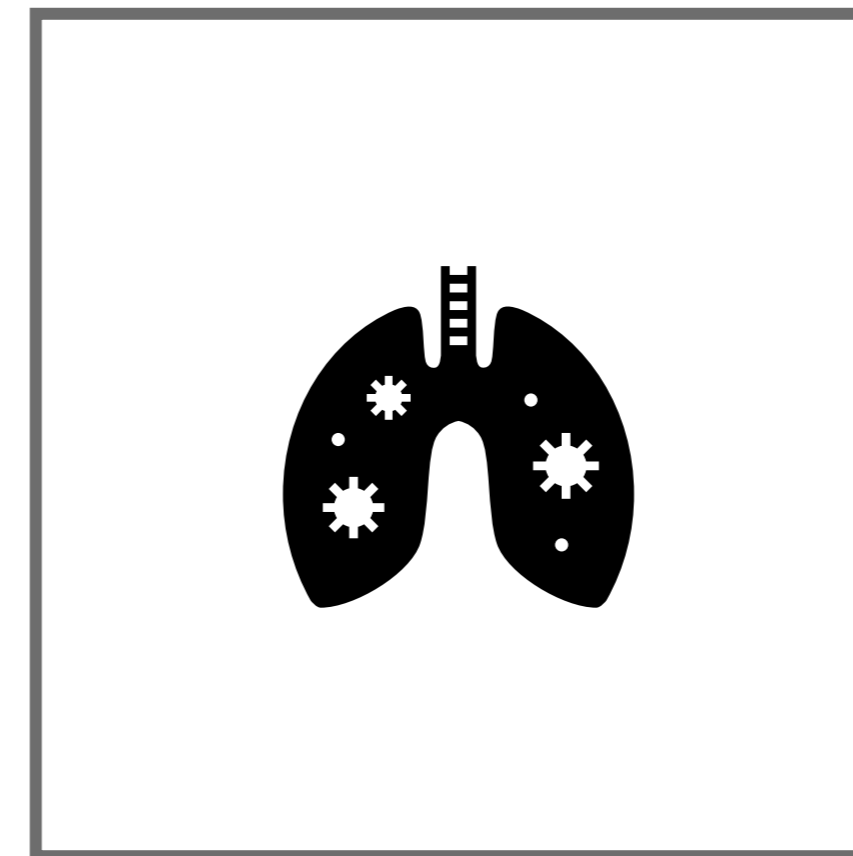
ARTIFICIAL INTELLIGENCE IDENTIFIES THOSE AT RISK



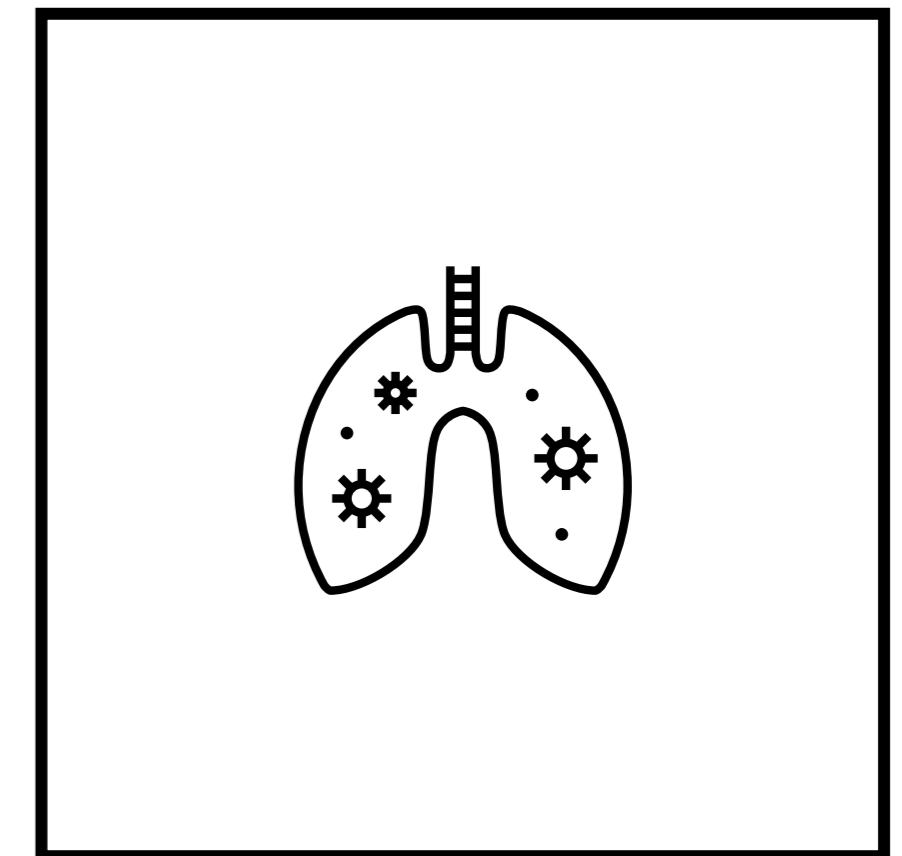
**EMR
DATA MINING**



**CT REPORT
DATA MINING**



**COMPUTER
ASSISTED
NODULE DETECTION**



**PATHWAY
SUGGESTION**

IMPROVES RECRUITMENT FOR EARLY DETECTION



DATA MINING



ARTIFICIAL INTELLIGENCE IDENTIFIES THOSE AT RISK

THE LANCET
Oncology

REVIEW · Volume 25, Issue 12, E694-E703, December 2024

Artificial intelligence-aided data mining of medical records for cancer detection and screening

[Amalie Dahl Haue, MD PhD^{a,b}](#) · [Jessica Xin Hjaltelin, PhD^a](#) · [Peter Christoffer Holm, PhD^a](#) · [Davide Placido, PhD^{a,b}](#) · [Søren Brunak, PhD^{a,b}](#)  

[Affiliations & Notes](#)  [Article Info](#) 

- a Novo Nordisk Foundation Center for Protein Research, Faculty of Health and Medical Sciences, University of Copenhagen, Copenhagen, Denmark
- b Copenhagen University Hospital Rigshospitalet, University of Copenhagen, Copenhagen, Denmark

In summary, the application of AI to multimodal, population-wide, health data brings the possibility of an array of methods for the mapping of malignant phenotypic spectra to a risk continuum. Therefore, avenues for developing cancer screening and early detection methods are wider than ever. The access to population-wide electronic health record data is an opportunity for predictive AI models to generalise to a much larger part of the heterogeneous phenotypic space of patients with cancer, and importantly, differentiate these patients from healthy controls that are also highly heterogeneous. Time will

IMPROVES RISK ASSESSMENT FOR EARLY DETECTION

ARTIFICIAL INTELLIGENCE IDENTIFIES THOSE AT RISK

JACC: ADVANCES

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VIEWPOINT

Population Health and Artificial Intelligence

R. Kannan Mutharasan, MD, MBA,^a Jessica Walradt, MS^b

Radiology: Artificial Intelligence

ORIGINAL RESEARCH




Assistive AI in Lung Cancer Screening: A Retrospective Multinational Study in the United States and Japan

Atilla P. Kiraly, PhD • Corbin A. Cunningham, PhD • Ryan Najafi, PhD • Zaid Nabulsi, MS • Jie Yang, PhD • Charles Lau, MD, MBA • Joseph R. Ledsam, MD • Wenxing Ye, PhD • Diego Ardila, MS • Scott M. McKinney, MS • Rory Pilgrim, BEng • Yun Liu, PhD • Hiroaki Saito, MD • Yasuteru Shimamura, MD • Mozziyar Etemadi, PhD, MD • David Melnick, MS • Sunny Jansen, PhD • Greg S. Corrado, PhD • Lily Peng, PhD, MD • Daniel Tse, MD • Shrayya Shetty, MS • Shruthi Prabhakara, PhD • David P. Naidich, MD • Neeral Beladia, MS • Krish Eswaran, PhD

From Google Health Research, 1600 Amphitheatre Pkwy, Mountain View, CA 94043 (A.P.K., C.A.C., R.N., Z.N., C.L., J.R.L., D.A., S.M.M., R.P., Y.L., S.J., G.S.C., L.P., D.T., S.S., S.P., K.E.); Waymo, Mountain View, Calif (J.Y., N.B.), David Geffen School of Medicine at UCLA, Los Angeles, Calif (C.L.); Google, Mountain View, Calif (W.Y.); Department of Gastroenterology, Sendai Kousei Hospital, Sendai, Japan (H.S.); MNES Inc, Hiroshima, Japan (Y.S.); Department of Telemedicine, Northwestern University Feinberg School of Medicine, Chicago, Ill (M.E., D.M.); and Center for Biological Imaging, New York University–Langone Medical Center, New York, NY (D.P.N.). Received March 23, 2023; revision requested May 22; revision received January 7, 2024; accepted March 1. **Address correspondence to** A.P.K. (email: akiraly@google.com).

Supported by Google.

Conflicts of interest are listed at the end of this article.

Radiology: Artificial Intelligence 2024; 6(3):e230079 • <https://doi.org/10.1148/ryai.230079> • Content codes:   

IMPROVES RECRUITMENT FOR EARLY DETECTION

WELLSTAR LCO: DATA MINING

NLP SCANS CT REPORTS FOR PRE-SELECTED TERMS AND PHRASES

OUR TEAM REVIEWS REPORTS FOR APPROPRIATE REFERRAL

IDENTIFIES PATIENTS AT RISK BY TERMINOLOGY

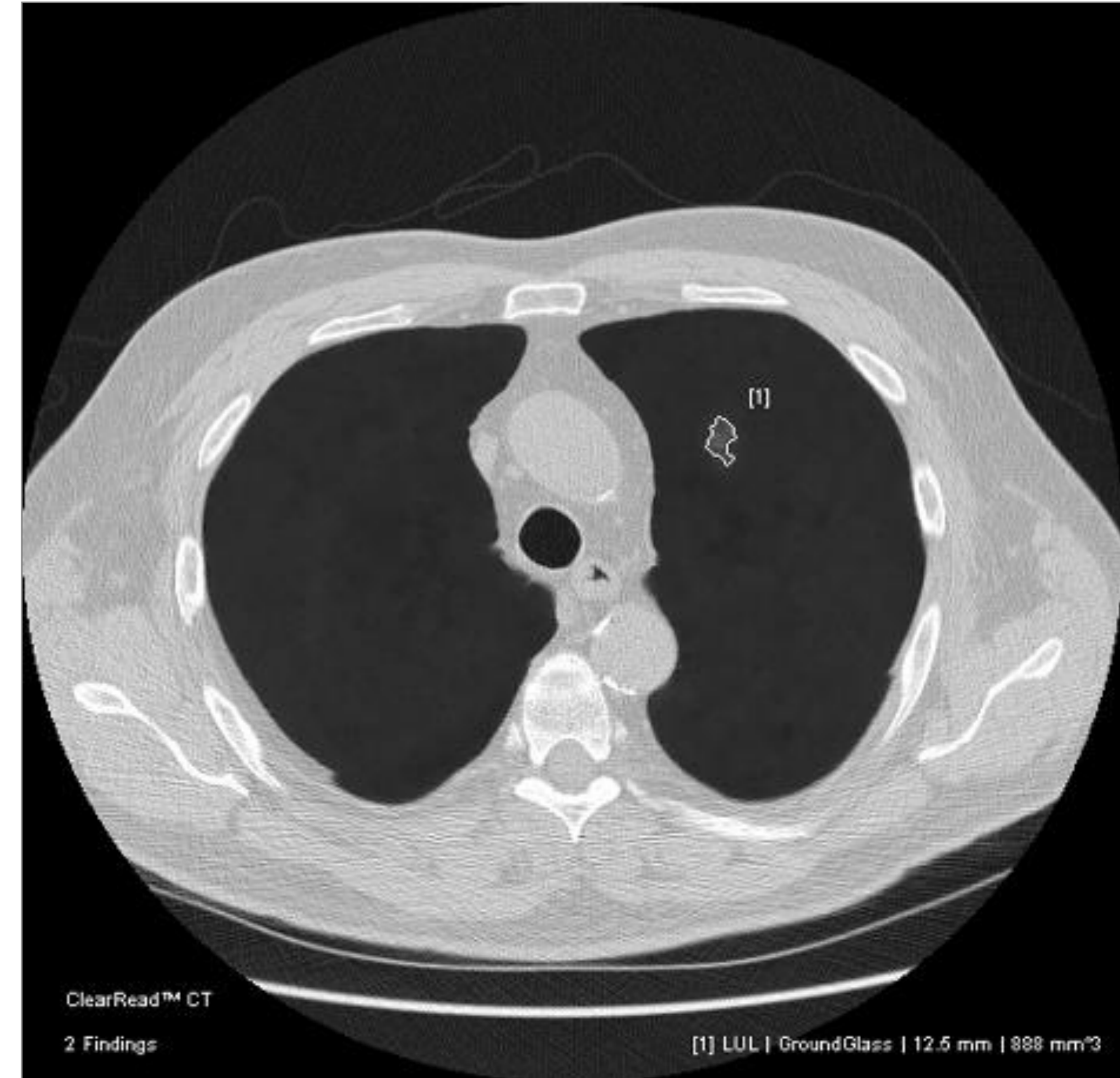
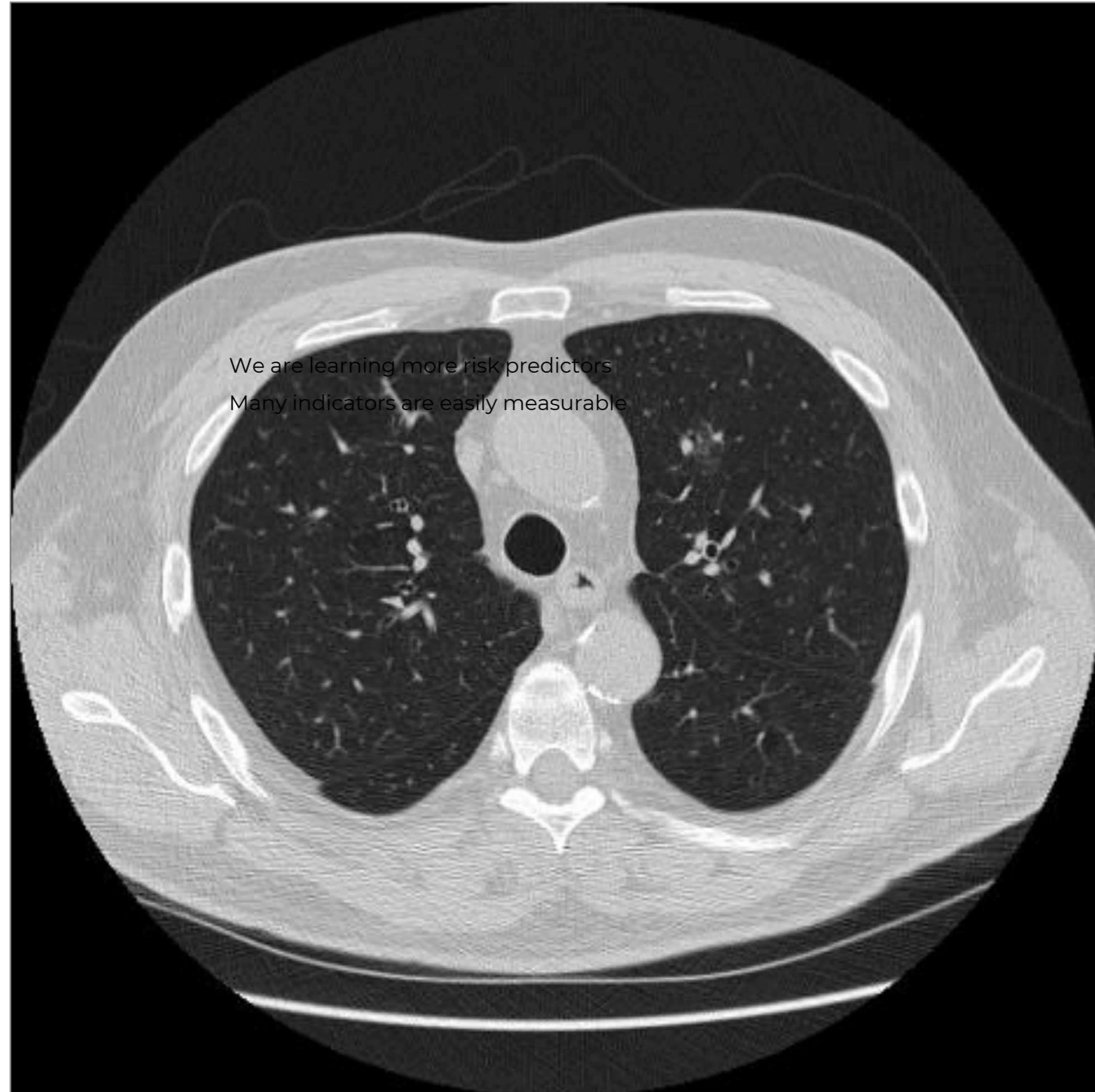
The screenshot displays the 'Lung Cancer Orchestrator' interface. At the top, there are navigation tabs for 'Screening', 'Incidental', and 'More', along with a dropdown menu for 'Medtronic LungGPS™ Platform'. Below this, there are filters for 'MRN', 'DOB', 'Gender', and 'Ethnicity'. The main content area shows a list of reports under the 'New Findings' tab. Two reports are visible, each with a 'Notice' and a 'Next Steps' panel. The first report's findings are highlighted in green: 'Multiple pulmonary nodules are demonstrated in the lung bases, for example there is a 6 mm nodule within the lingula on series 204 image 12. Linear opacities in the lung bases consistent with subsegmental atelectasis or scarring. There is incompletely imaged azygosastria.' The second report's findings are also highlighted: 'Lungs: Mild nodular biapical scarring is present. A trivial left apical pneumothorax is present, better seen on the comparison CT. Focal subpleural nodular opacity is present in the posterior segment of the right upper lobe measuring 22 x 15 mm (image 58 series 507). This is adjacent to right fifth and sixth rib fractures. There are several additional well-circumscribed solid noncalcified bilateral juxtapleural pulmonary nodules, most compatible with benign intrapulmonary lymph nodes.' The 'Next Steps' panel for the second report shows 'Defer: ILN' as a selected action.

RIVERAIN

NODULE DETECTION

FDA APPROVED

AI REMOVES
VASCULATURE
TO REVEAL
NODULES



RIVERAIN

NODULE DETECTION REPORT CREATION

AI CREATES A
**SUMMARY
REPORT**


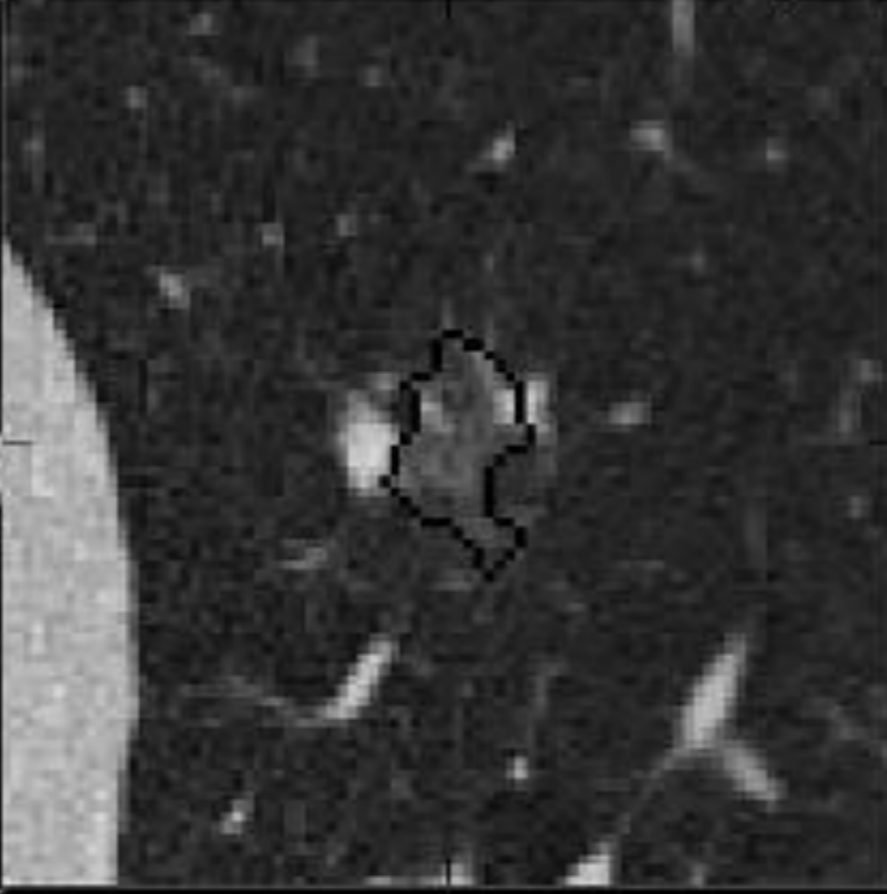
**INSERTS INTO
POWERSCRIBE**

Accession D015-00002
Study Date 2017-11-17
Series # 485

ClearRead CT Detect Summary Report

[1] Image # 68

Ground Glass LUL

Nodule	
Avg Diameter	12.5 mm
Min Diameter	9.3 mm
Max Diameter	15.6 mm
Z-Diameter	13.7 mm
Volume	888 mm ³

2 Findings

[1] Image # 68

Ground Glass LUL

Avg Diameter 12.5 mm
Volume 888 mm³

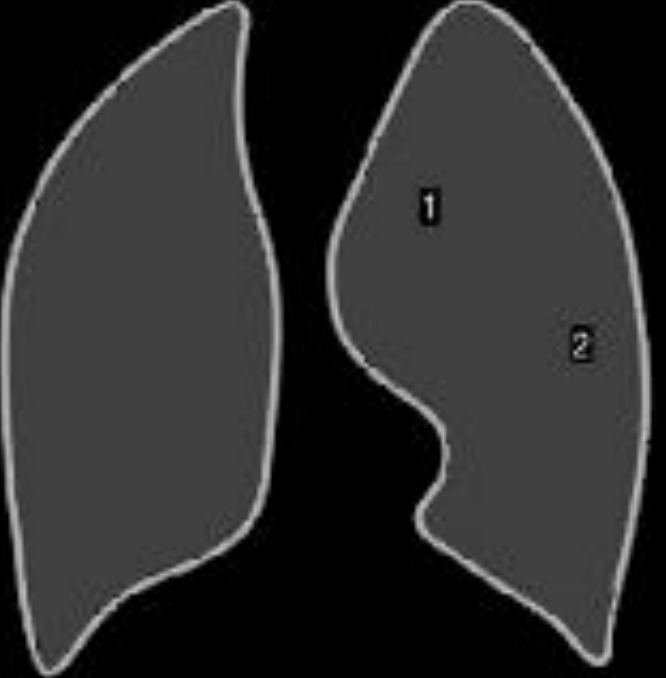
[2] Image # 113

Solid LLL

Avg Diameter 4.6 mm
Volume 64 mm³

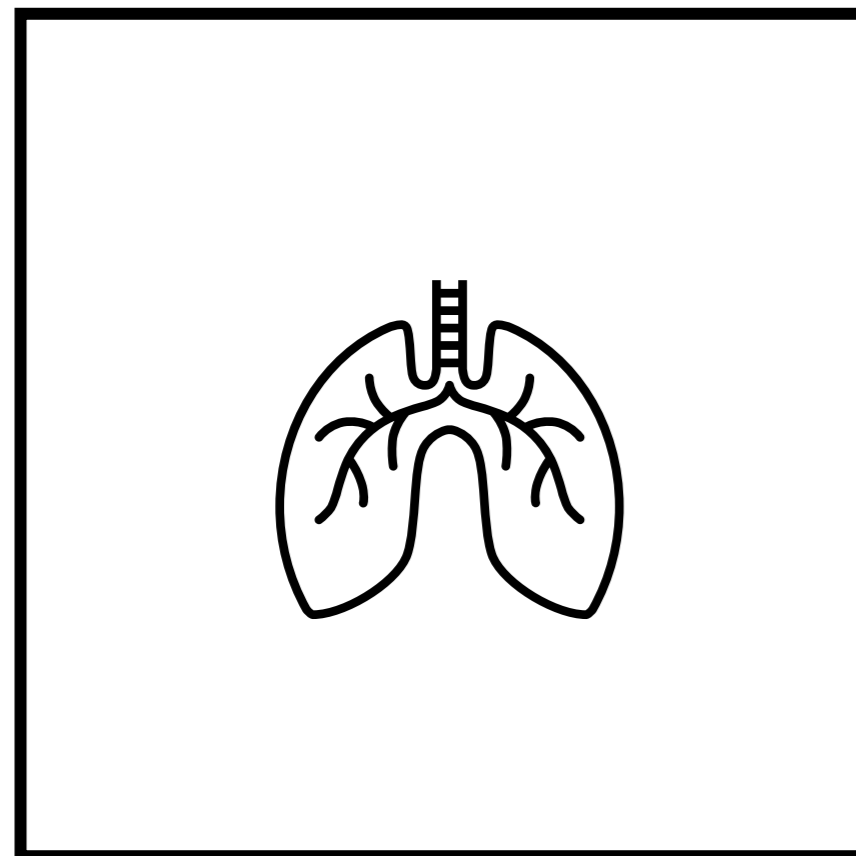
ClearRead CT Detect Summary Report

Accession D015-00002
Study Date 2017-11-17
Series # 485



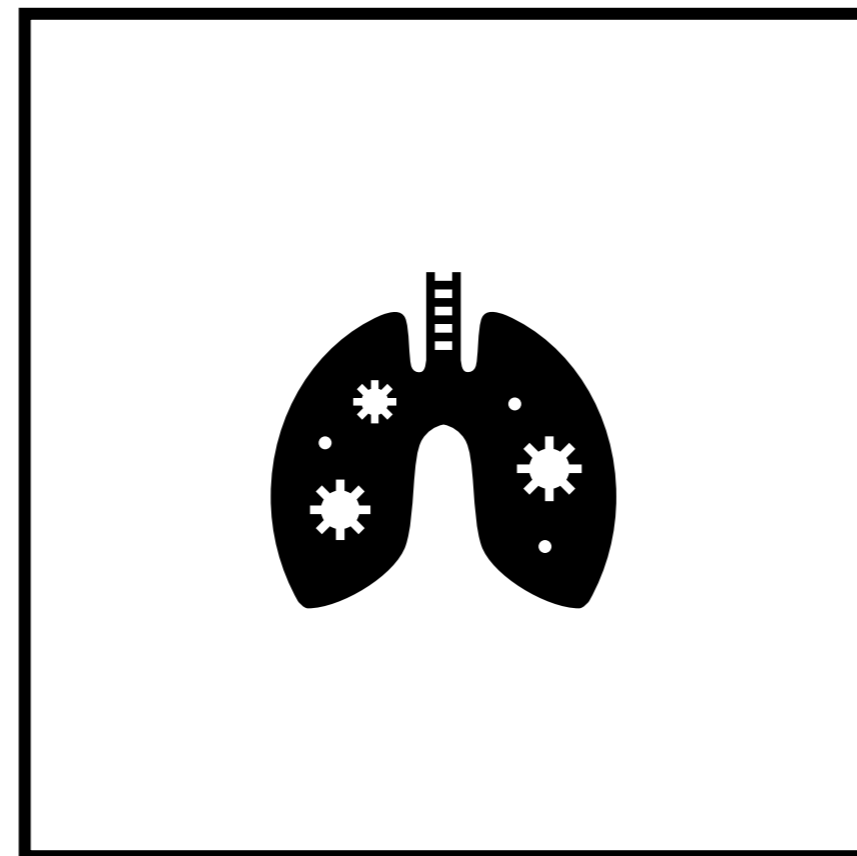
	Right	Left
Volume	3.10L	2.70L
# Findings	0	2
Largest	NA	1(12.5mm)

ARTIFICIAL INTELLIGENCE: IDENTIFY



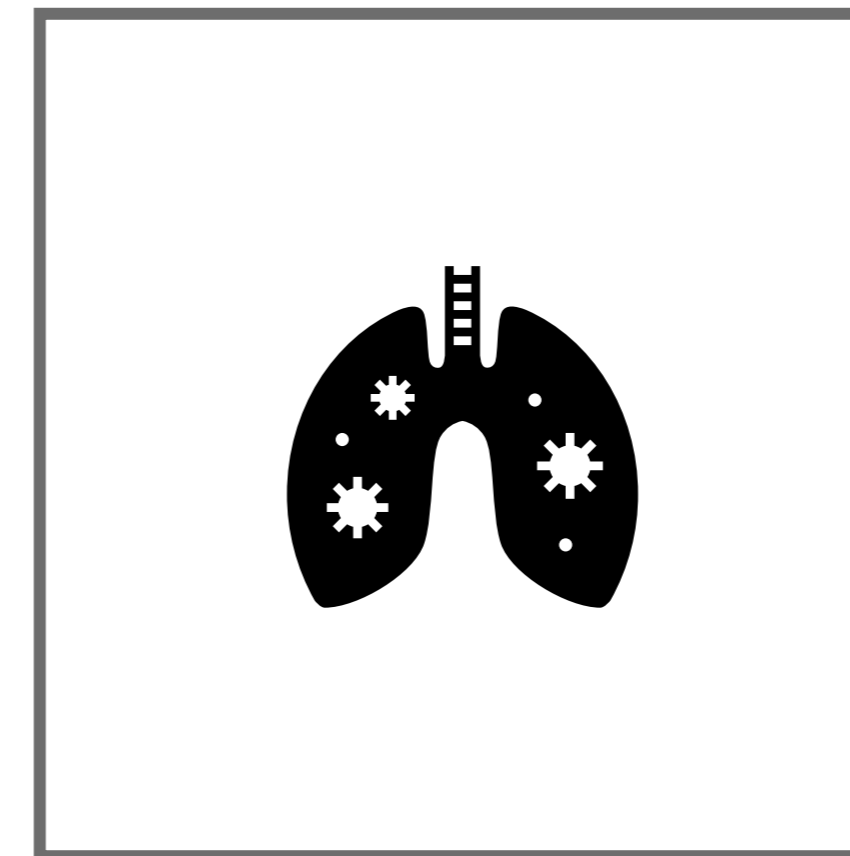
AZRA AI

AI INTERPRETS
UNSTRUCTURED DATA



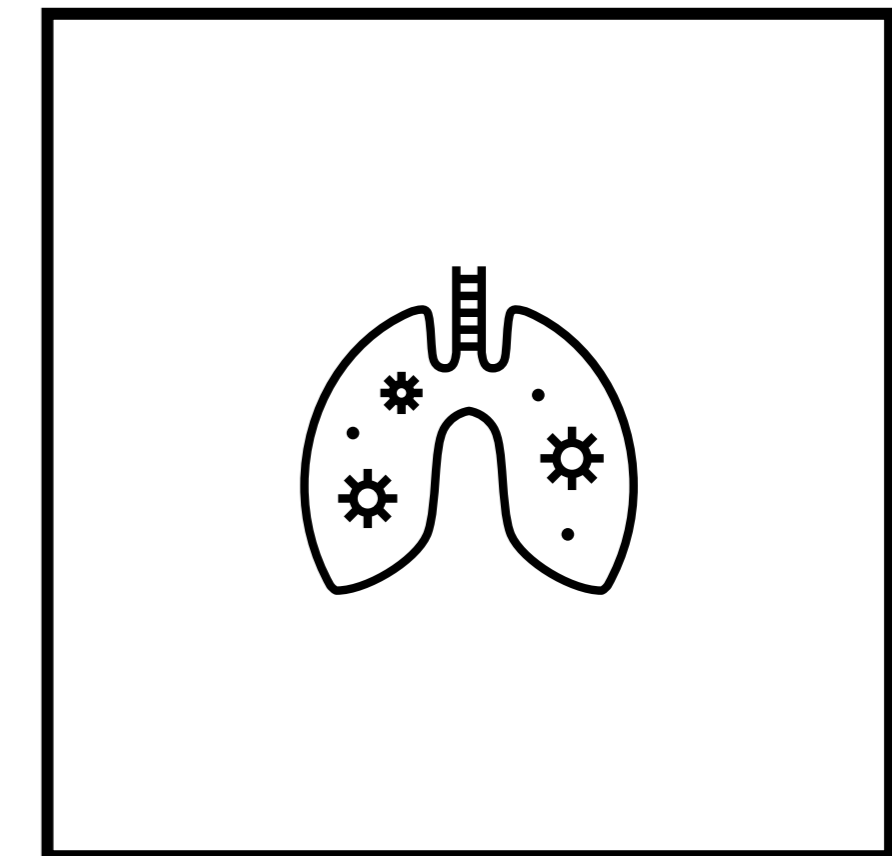
EON AZRA AI

READS CT SCAN REPORTS
IDENTIFIES RISK CRITERIA



LUNG CANCER ORCHESTRATOR

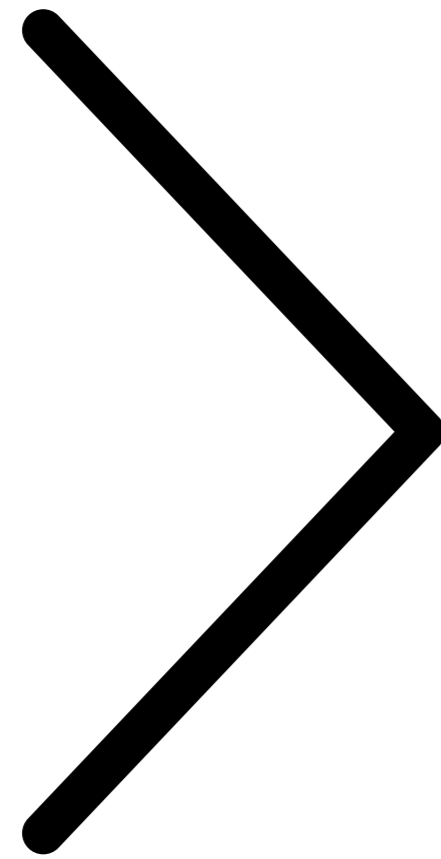
SCANS LDCT REPORTS
FOR KEY WORDS AND
PHRASES



RIVERAIN

**NODULE DETECTION
CREATES NODULE
TABLE,**

LOOK-BACK



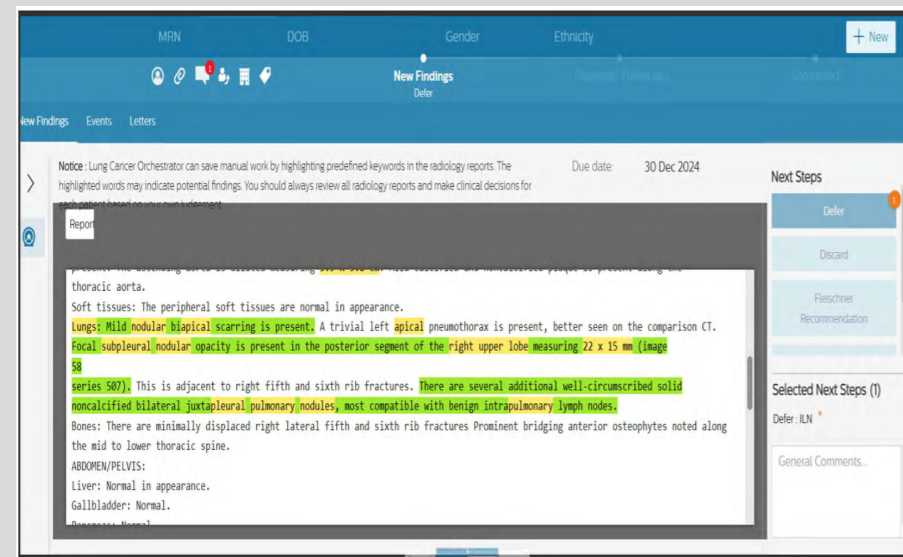
EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

STRATIFY RISK

SEPARATE LOW RISK FROM HIGH RISK

STRATIFY RISK

SEPARATE HIGHER RISK FROM LOWER RISK .



BY CLINICAL FACTORS

AI STRATIFIES DATA MINING RESULTS

IDENTIFY THOSE AT HIGHER RISK

BY TERMINOLOGY

NODULES DESCRIBED OVER 8 MM ARE PRIORITIZED



BY NODULE SIZE ON CT

NODULE TABLE SORTS NODULES BY SIZE

BY NODULE CHARACTER

BORDER ANALYSIS, SOLID, SEMI-SOLID, GGO, GROWTH

STRATIFY RISK

SEPARATE HIGHER RISK FROM LOWER RISK .

EON

Without Eon

```

MS-DOS Prompt
8 x 12
-----
FirstName Bradley
LastName High
DOB 10/9/65
Abnormality Lesion
Size 4.5cm
            
```

VS.

With Eon

eon

HIGH, BRADLEY DOB: OCTOBER 9, 1965 AGE: 59

Eon Computational Linguistics Fields

CYST FINDINGS

Abnormality	Size	Ductal Dialation	Location
Lesion	4.5 cm diameter	Dialation	Pancreatic Ductal Biliary
Characterization	Location	Size	
Cystic	Head	5 mm diameter	
IPMN	Uncinate		
Complex	Process		
Multi-loculated			
Simple			

ATS American Journal of Respiratory and Critical Care Medicine

Assessing the Accuracy of a Deep Learning Method to Risk Stratify Indeterminate Pulmonary Nodules

Pierre P. Massion^{1,2}, Sanja Antic¹, Sarim Ather³, Carlos Arteta⁴, Jan Brabec⁵, Heidi Chen⁴, Jerome Declerck³, David Dufek⁵, William Hickey³, Timor Kadir⁴, Jonas Kunst⁵, Bennett A. Landman⁷, Reginald F. Munden⁸, Petr Novotny⁴, Heiko Peschl³, Lyndsey C. Pickup⁴, Catarina Santos⁴, Gary T. Smith^{9,10}, Ambika Talwar³, and Fergus Gleeson³

<http://www.atsjournals.org/doi/abs/10.1164/rccm.201503-0511AC>

ABSTRACT

Rationale: The management of indeterminate pulmonary nodules (IPNs) remains challenging, resulting in invasive procedures and delays in diagnosis and treatment. Strategies to decrease the rate of unnecessary invasive procedures and optimize surveillance regimens are needed.

Objectives: To develop and validate a deep learning method to improve the management of IPNs.

Methods: A Lung Cancer Prediction Convolutional Neural Network model was trained using computed tomography images of IPNs from the National Lung Screening Trial, internally validated, and externally tested on cohorts from two academic institutions.

Measurements and Main Results: The areas under the receiver operating characteristic curve in the external validation cohorts were 83.5% (95% confidence interval [CI], 75.4–90.7%) and 91.9% (95% CI, 88.7–94.7%), compared with 78.1% (95% CI, 68.7–86.4%) and 81.9 (95% CI, 76.1–87.1%), respectively, for a commonly used clinical risk model for incidental nodules. Using 5% and 65% malignancy thresholds defining low- and high-risk categories, the overall net reclassifications in the validation cohorts for cancers and benign nodules compared with the Mayo model were 0.34 (Vanderbilt) and 0.30 (Oxford) as a rule-in test, and 0.33 (Vanderbilt) and 0.58 (Oxford) as a rule-out test. Compared with traditional risk prediction models, the Lung Cancer Prediction Convolutional Neural Network was associated with

improved accuracy in predicting the likelihood of disease at each threshold of management and in our external validation cohorts.

Conclusions: This study demonstrates that this deep learning algorithm can correctly reclassify IPNs into low- or high-risk categories in more than a third of cancers and benign nodules when compared with conventional risk models, potentially reducing the number of unnecessary invasive procedures and delays in diagnosis.

Keywords: early detection; risk stratification; neural networks; lung cancer; computer-aided image analysis

STRATIFY RISK

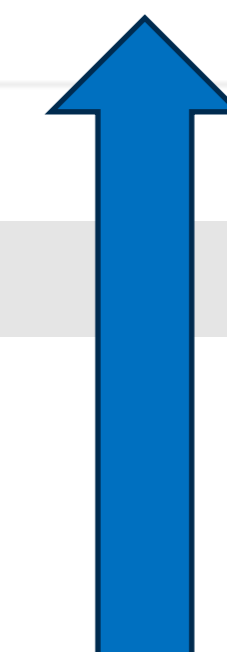
SEPARATE HIGHER RISK FROM LOWER RISK .

WELLSTAR LCO



Nodule size is listed in LCO worklist which helps the coordinators prioritize working their patients.

Name	MRN	Gender	Age	Creation Date	Status	Sub-status	Defer/Discard Reason	Unreviewed report	Due date	Notes	Assignee	Managed By Facility	Tags	Actions
		Male	42	05 Oct 2024	New Findings			0	15 Oct 2024			WellstarMF_KH	1-2 mm	
		Male	40	05 Oct 2024	New Findings			0	15 Oct 2024			WellstarMF_KH	7 x 5 mm	
		Female	89	05 Oct 2024	New Findings			0	15 Oct 2024			WellstarMF_NF	4 mm	
		Female	65	06 Oct 2024	New Findings			1	16 Oct 2024			WellstarMF_DH		
		Female	83	06 Oct 2024	New Findings			0	16 Oct 2024			SP		



RIVERAIN

STRATIFY RISK

Automatic
 nodule
 comparison
 from current
 and prior
 exam

5 Findings
 User responsible for confirming nodule matches

ClearRead CT Compare Summary Report

Accession D015-00008
 Study Date 1998-03-03
 Series # 5418

Prior Accession D015-00008
 Prior Study Date 1998-01-01
 Prior Series # 3163

Image #	Prior #	Location	Avg Diameter	Volume	% Change
[1] Image # 12	Prior # 15	Solid RUL 166d	7.9 mm << 6.6 mm	463 mm ³ << 359 mm ³	+19% / +29%
[2] Image # 27	Prior # 29	Solid LUL 398d	20.7 mm << 22.1 mm	5014 mm ³ << 4509 mm ³	-6% / +11%
[3] Image # 32	Prior # 34	Solid RLL 95d	5.4 mm << 7.3 mm	472 mm ³ << 303 mm ³	-26% / +56%
[4] Image # 40	-	Solid RUL	11.1 mm	1094 mm ³	-
[5] Image # 77	Prior # 79	Solid LLL 27d	8.5 mm << 4.6 mm	261 mm ³ << 55 mm ³	+84% / +375%

No match

Page 1 of 6

ClearRead CT Compare Summary Report

Accession D015-00008
 Study Date 1998-03-03
 Series # 5418

Prior Accession D015-00008
 Prior Study Date 1998-01-01
 Prior Series # 3163

Image #	Prior #	Location	Avg Diameter	Volume	% Change
[5] Current Image # 77	Prior Image # 79	Solid LLL 27d	8.5 mm	261 mm ³	+84% / +375%
-	-	Nodule	4.6 mm	55 mm ³	-

Volume Change

User responsible for confirming nodule matches

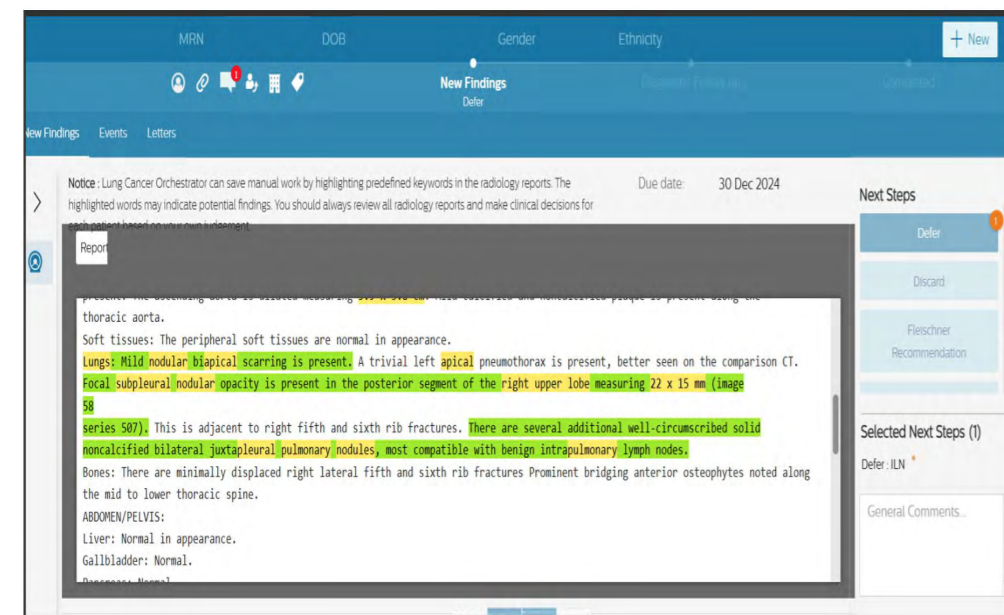
5 Findings

Page 6 of 6

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

STRATIFY RISK

SEPARATE HIGHER RISK FROM LOWER RISK.



BY CLINICAL FACTORS

AZRA AI

EON

LUNG CANCER ORCHESTRATOR

BY TERMINOLOGY

AZRA AI

EON

LUNG CANCER ORCHESTRATOR

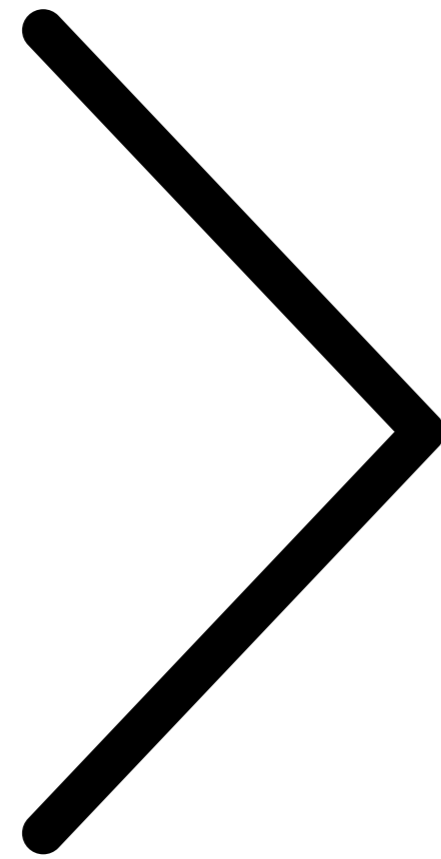
BY NODULE SIZE

EON

LUNG CANCER ORCHESTRATOR

BY NODULE CHARACTER

GGO, NON-SOLID, SOLID,
SPICULATED



EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

ANALYZE

RISK

WHO SHOULD GET A WORKUP

External validation of a convolutional neural network artificial intelligence tool to predict malignancy in pulmonary nodules

David R Baldwin,¹ Jennifer Gustafson,² Lyndsey Pickup,³ Carlos Arteta,³ Petr Novotny,⁴ Jerome Declerck,³ Timor Kadir,³ Catarina Figueiras,² Albert Sterba,⁵ Alan Exell,⁶ Vaclav Potesil,³ Paul Holland,⁷ Hazel Spence,⁷ Alison Clubley,⁷ Emma O'Dowd,¹ Matthew Clark,⁸ Victoria Ashford-Turner,⁹ Matthew EJ Callister,⁹ Fergus V Gleeson²

ABSTRACT

<https://thorax.bmj.com/content/75/4/304.full.pdf>

Background: Estimation of the risk of malignancy in pulmonary nodules detected by CT is central in clinical management. The use of artificial intelligence (AI) offers an opportunity to improve risk prediction. Here we compare the performance of an AI algorithm, the lung cancer prediction convolutional neural network (LCP-CNN), with that of the Brock University model, recommended in UK guidelines.

Methods: A dataset of incidentally detected pulmonary nodules measuring 5–15 mm was collected retrospectively from three UK hospitals for use in a validation study. Ground truth diagnosis for

each nodule was based on histology (required for any cancer), resolution, stability or (for pulmonary lymph nodes only) expert opinion. There were 1397 nodules in 1187 patients, of which 234 nodules in 229 (19.3%) patients were cancer. Model discrimination and performance statistics at predefined score thresholds were compared between the Brock model and the LCP-CNN.

Results: The area under the curve for LCP-CNN was 89.6% (95% CI 87.6 to 91.5), compared with 86.8% (95% CI 84.3 to 89.1) for the Brock model ($p < 0.005$). Using the LCP-CNN, we found that 24.5% of nodules

scored below the lowest cancer nodule score, compared with 10.9% using the Brock score. Using the predefined thresholds, we found that the LCP-CNN gave one false negative (0.4% of cancers), whereas the Brock model gave six (2.5%), while specificity statistics were similar between the two models.

Conclusion: The LCP-CNN score has better discrimination and allows a larger proportion of benign nodules to be identified without missing cancers than the Brock model. This has the potential to substantially reduce the proportion of surveillance CT scans required and thus save significant resources.

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

NODULE ANALYSIS BY AI

OPTELLUM

FDA APPROVED

Artificial Intelligence Tool for Assessment of Indeterminate Pulmonary Nodules Detected with CT

Roger Y. Kim, Jason L. Oke, Lyndsey C. Pickup, Reginald F. Munden, Travis L. Dotson, Christina R. Bellinger, Avi Cohen, Michael J. Simoff, Pierre P. Massion, Claire Filippini, Fergus V. Gleeson, Anil Vachani

ABSTRACT

<https://pubs.rsna.org/doi/epdf/10.1148/radiol.20211212>

Background: Limited data are available regarding whether computer-aided diagnosis (CAD) improves assessment of malignancy risk in indeterminate pulmonary nodules (IPNs).

Purpose: To evaluate the effect of an artificial intelligence-based CAD tool on clinician IPN diagnostic performance and agreement for both malignancy risk categories and management recommendations.

Materials and Methods: This was a retrospective multireader multicase study performed in June and July 2020 on chest CT studies of IPNs. Readers used only CT imaging data and provided an estimate of malignancy risk and a management recommendation for each case without and with CAD. The effect of CAD on average reader diagnostic performance

was assessed using the Obuchowski-Rockette and Dorfman-Berbaum-Metz method to calculate estimates of area under the receiver operating characteristic curve (AUC), sensitivity, and specificity. Multirater Fleiss κ statistics were used to measure interobserver agreement for malignancy risk and management recommendations.

Results: A total of 300 chest CT scans of IPNs with maximal diameters of 5–30 mm (50.0% malignant) were reviewed by 12 readers (six radiologists, six pulmonologists) (patient median age, 65 years; IQR, 59–71 years; 164 [55%] men). Readers' average AUC improved from 0.82 to 0.89 with CAD ($P < .001$). At malignancy risk thresholds of 5% and 65%, use of CAD improved average sensitivity from 94.1% to 97.9% ($P = .01$) and from 52.6% to 63.1% ($P < .001$),

respectively. Average reader specificity improved from 37.4% to 42.3% ($P = .03$) and from 87.3% to 89.9% ($P = .05$), respectively. Reader interobserver agreement improved with CAD for both the less than 5% (Fleiss κ , 0.50 vs 0.71; $P < .001$) and more than 65% (Fleiss κ , 0.54 vs 0.71; $P < .001$) malignancy risk categories. Overall reader interobserver agreement for management recommendation categories (no action, CT surveillance, diagnostic procedure) also improved with CAD (Fleiss κ , 0.44 vs 0.52; $P = .001$).

Conclusion: Use of computer-aided diagnosis improved estimation of indeterminate pulmonary nodule malignancy risk on chest CT scans and improved interobserver agreement for both risk stratification and management recommendations.

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER.

NODULE ANALYSIS BY AI

OPTELLUM
FDA APPROVED

The screenshot displays the Optellum LCP Score interface. At the top, a timeline shows a study on Jul 2 with a score of 10 for RML_1. The main area shows a CT scan of the chest with a nodule highlighted in purple, labeled 'RML_1 (Solid) Score: 10'. To the right, a radiology report is shown in draft status, dated Mar 9, 2022. The report includes clinical information, findings, and a description of the nodule. Below the report, the Mayo score (19.4%) and Brock score (10.42%) are displayed. On the far right, a bar chart titled 'Optellum LCP Score' shows the percentage of cancer for scores 1 through 10. A warning box indicates that the patient population's cancer prevalence must be considered when interpreting the score. At the bottom right, there are buttons for 'Discharge', 'Refer to lung cancer team', 'Flag for attention', and 'Booking'.

Optellum LCP Score

Highest Score (latest study): 10 RML_1

Optellum LCP score	% cancer
1	0.2%
2	0.4%
3	0.8%
4	2.0%
5	5.6%
6	15%
7	34%
8	64%
9	84%
10	93%

Clinical decision

Last decision: none

Please make a follow-up decision for this patient.

Follow-up decision:

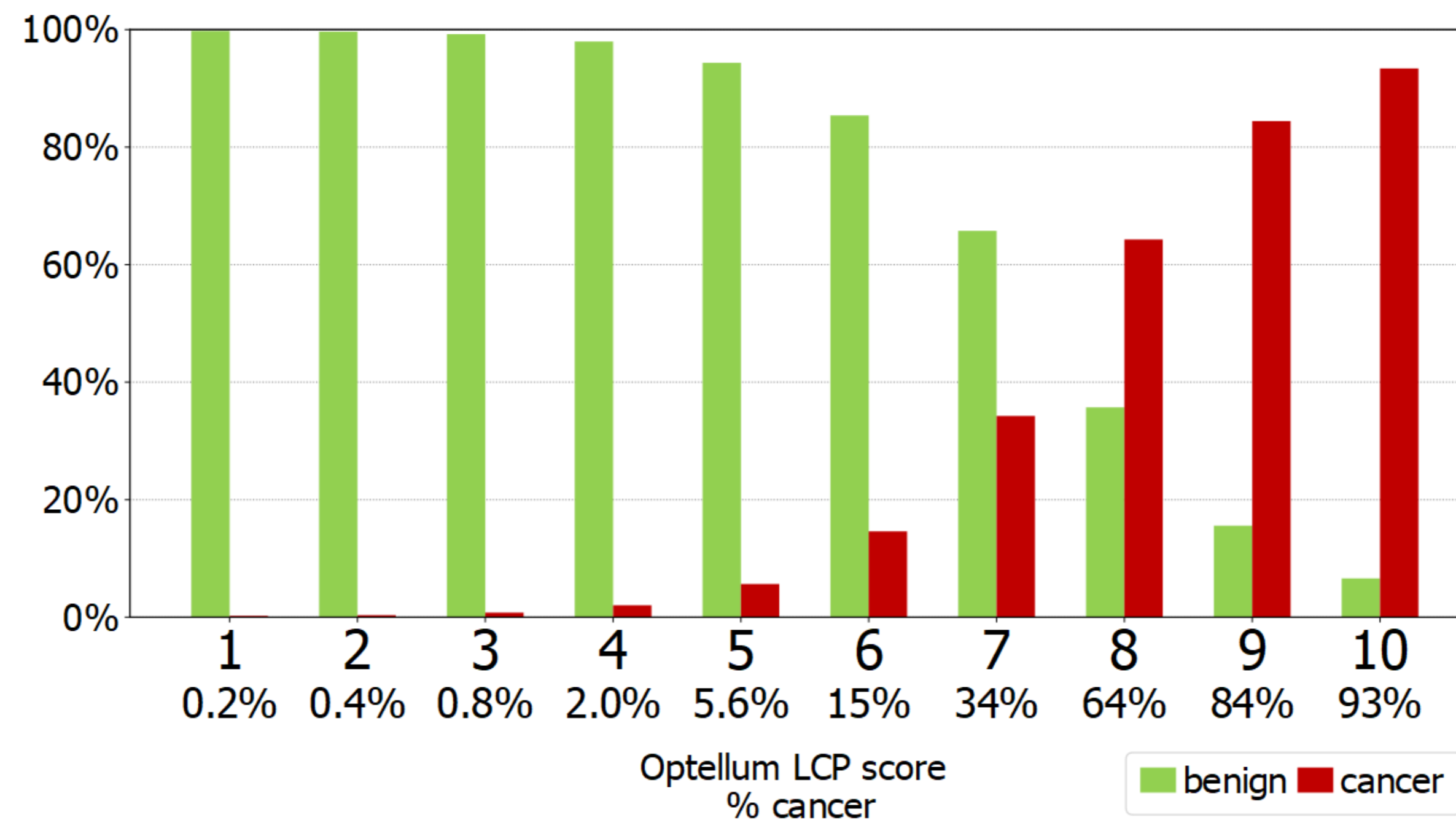
Flag for attention

Booking

No booking date to update.

Optellum Lung Cancer Prediction (LCP) score

- The only FDA-cleared AI/Radiomics malignancy biomarker for incidentally detected nodules¹⁻³
- Nodule malignancy score from 1 (benign) to 10 (malignant), based on neural network analysis of a standard CT
- Reimbursed by Medicare (CMS New Technology APC, \$650/patient)
- Helps prioritize and guide the right patients/nodules into invasive procedures⁴
- Integrated into the Virtual Nodule Clinic platform - WellStar the first user in Georgia



[1] Massion, Pierre P., et al. "Assessing the accuracy of a deep learning method to risk stratify indeterminate pulmonary nodules." *AJRCCM* 202.2 (2020): 241-249.

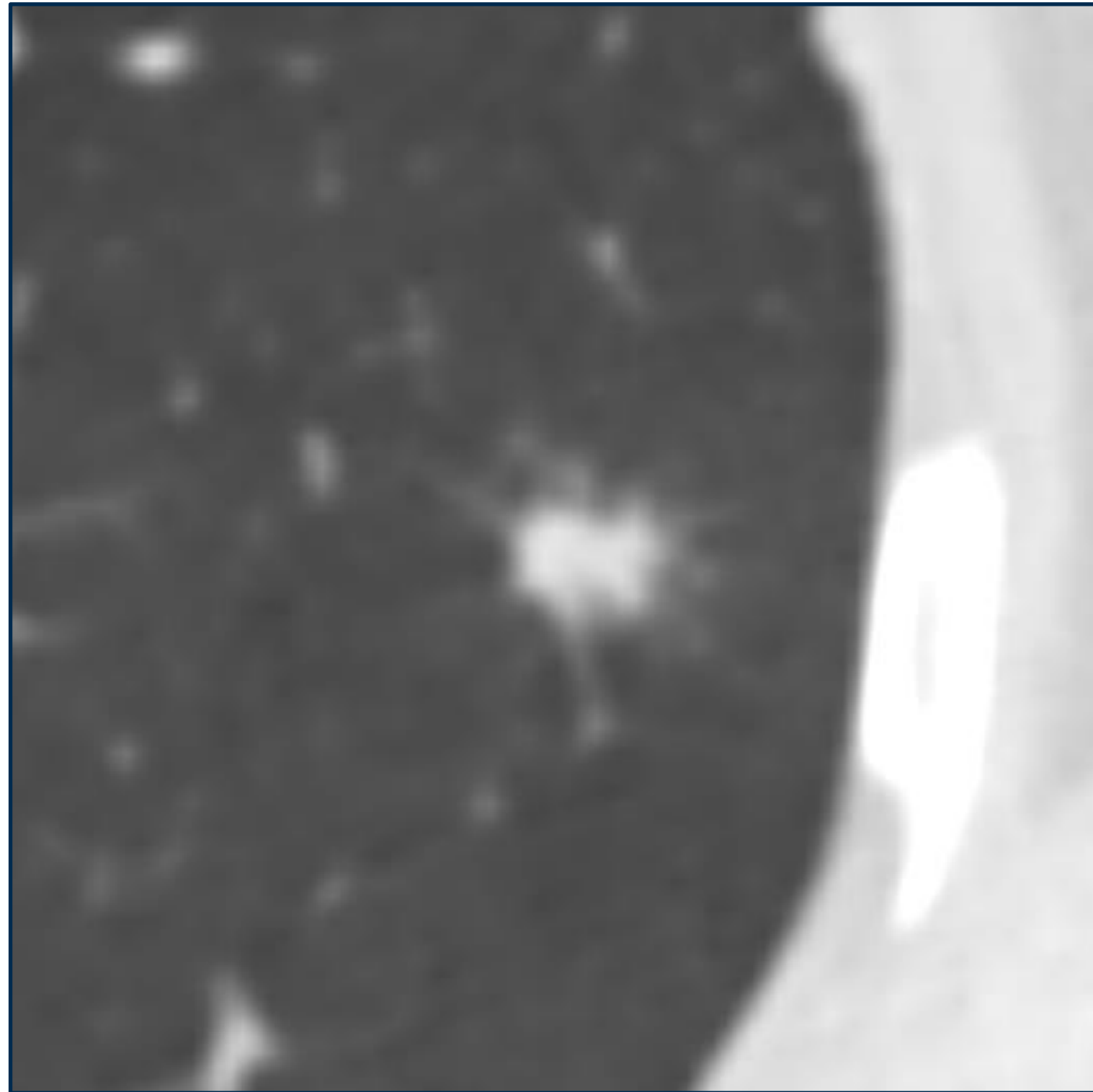
[2] Baldwin, David R., et al. "External validation of a convolutional neural network artificial intelligence tool to predict malignancy in pulmonary nodules." *Thorax* 75.4 (2020): 306-312.

[3] Kim, Roger Y., et al. Kim, Roger Y., et al. "Artificial intelligence tool for assessment of indeterminate pulmonary nodules detected with CT." *Radiology* 304.3 (2022): 683-691.

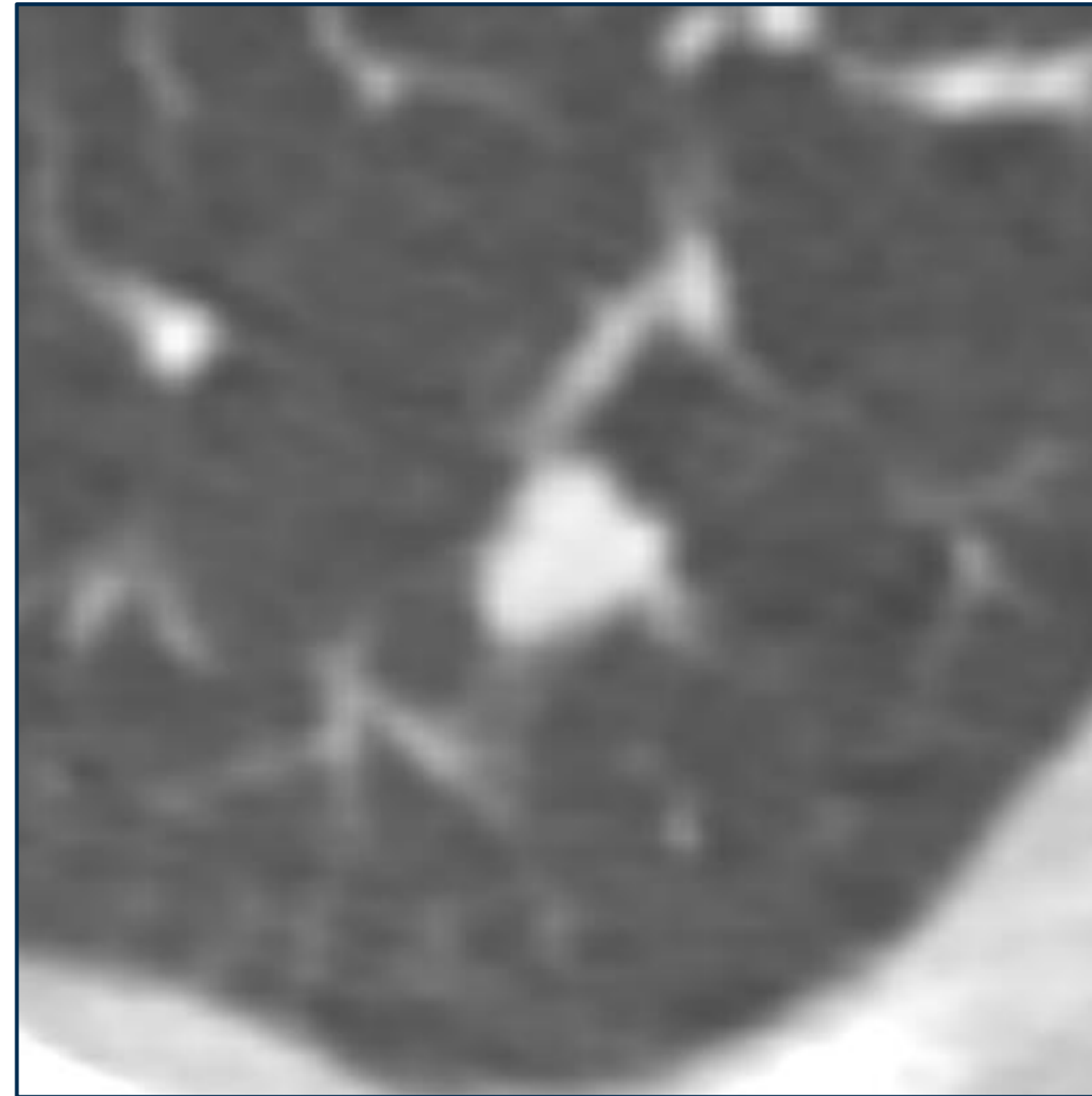
[4] Miller, Daniel., et al. "Clinical Application of an Artificial Intelligence Software for Prediction of Lung Cancer." STSA (2024)

Examples of malignant nodules

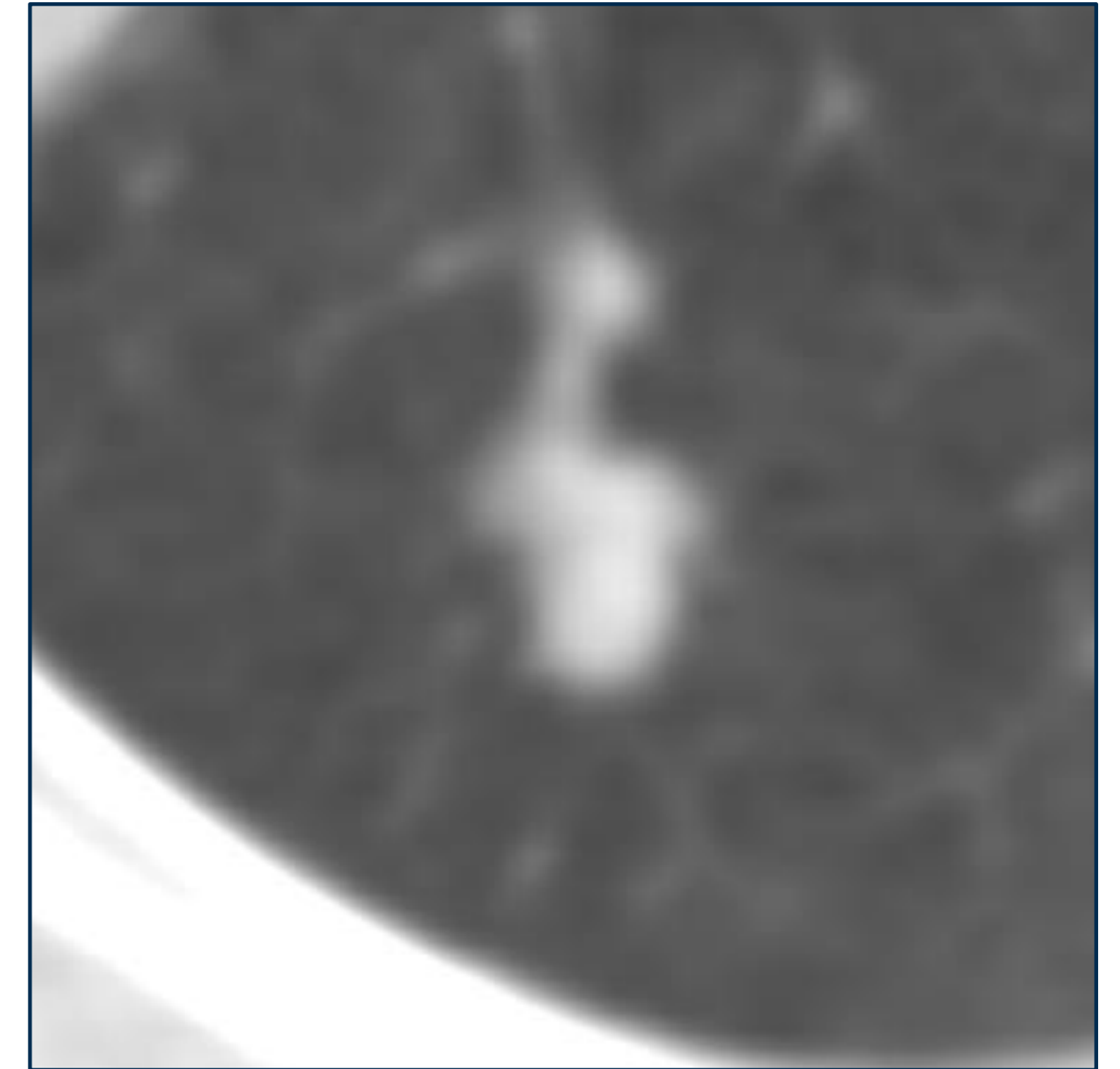
10mm



68yo M
8mm LUL
LCP score: 8
Adenocarcinoma



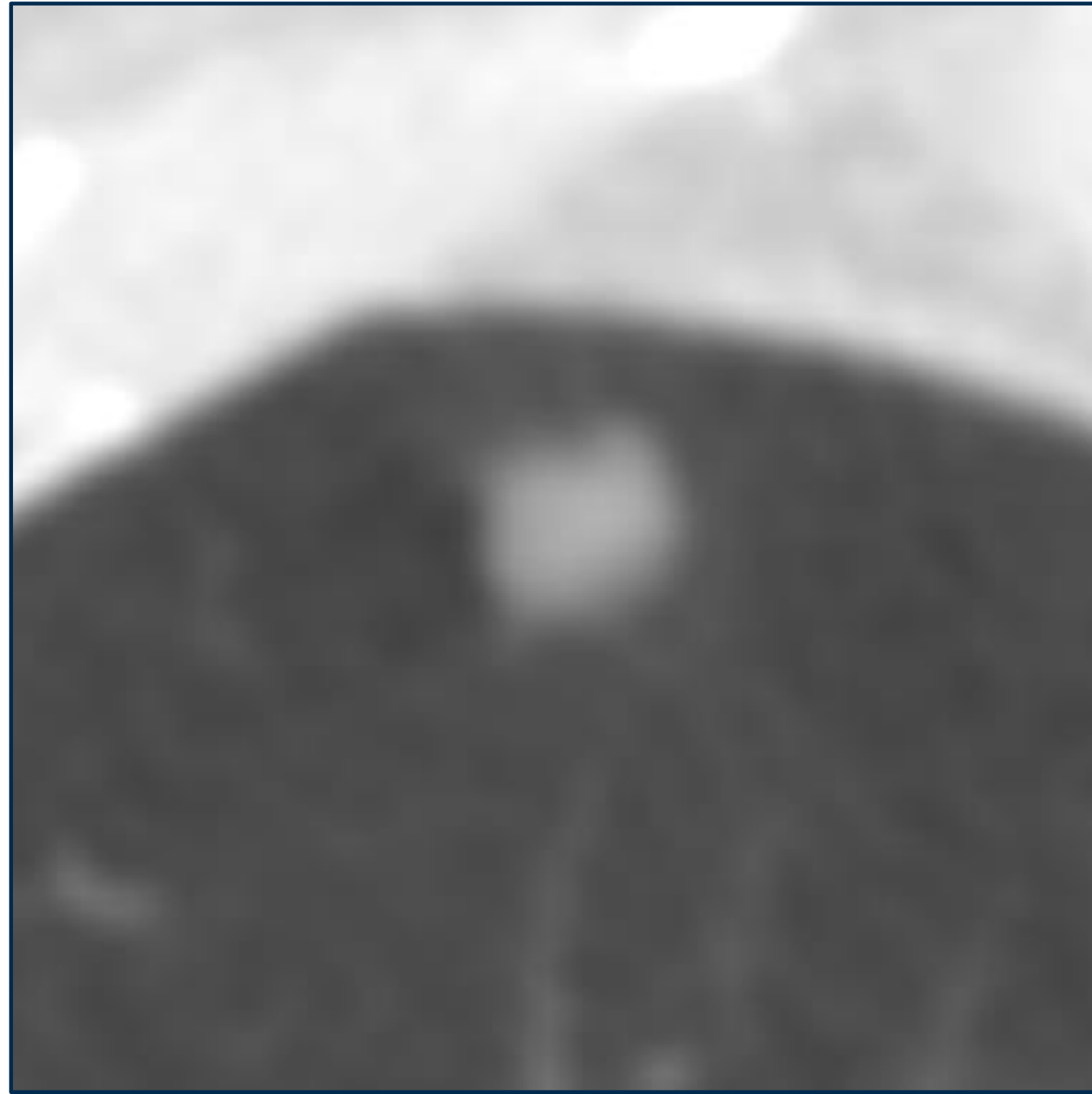
56yo F
8mm RUL
LCP score: 8
Adenocarcinoma



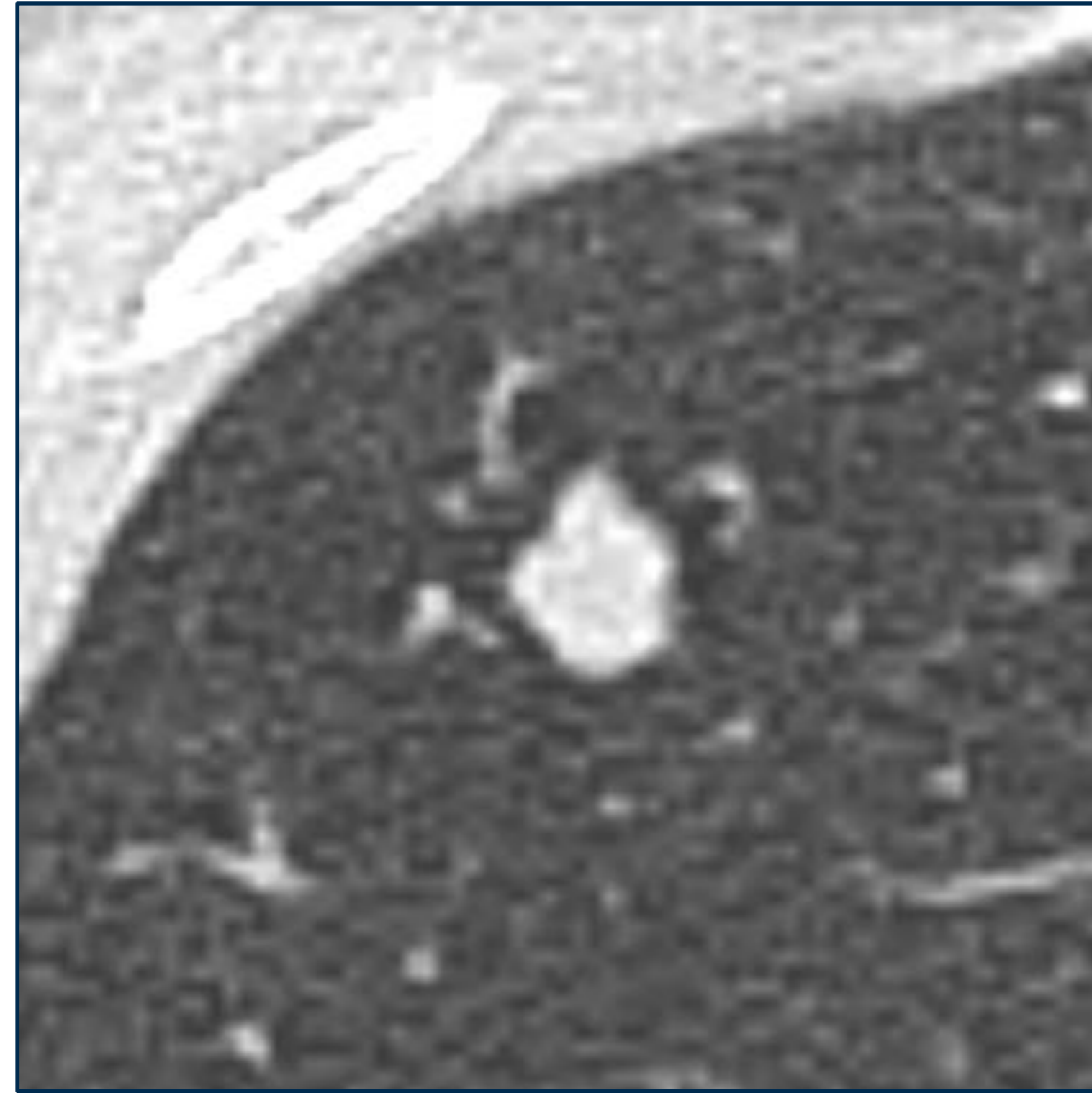
61yo F
10mm LLL
LCP score: 9
Adenocarcinoma

Examples of benign nodules

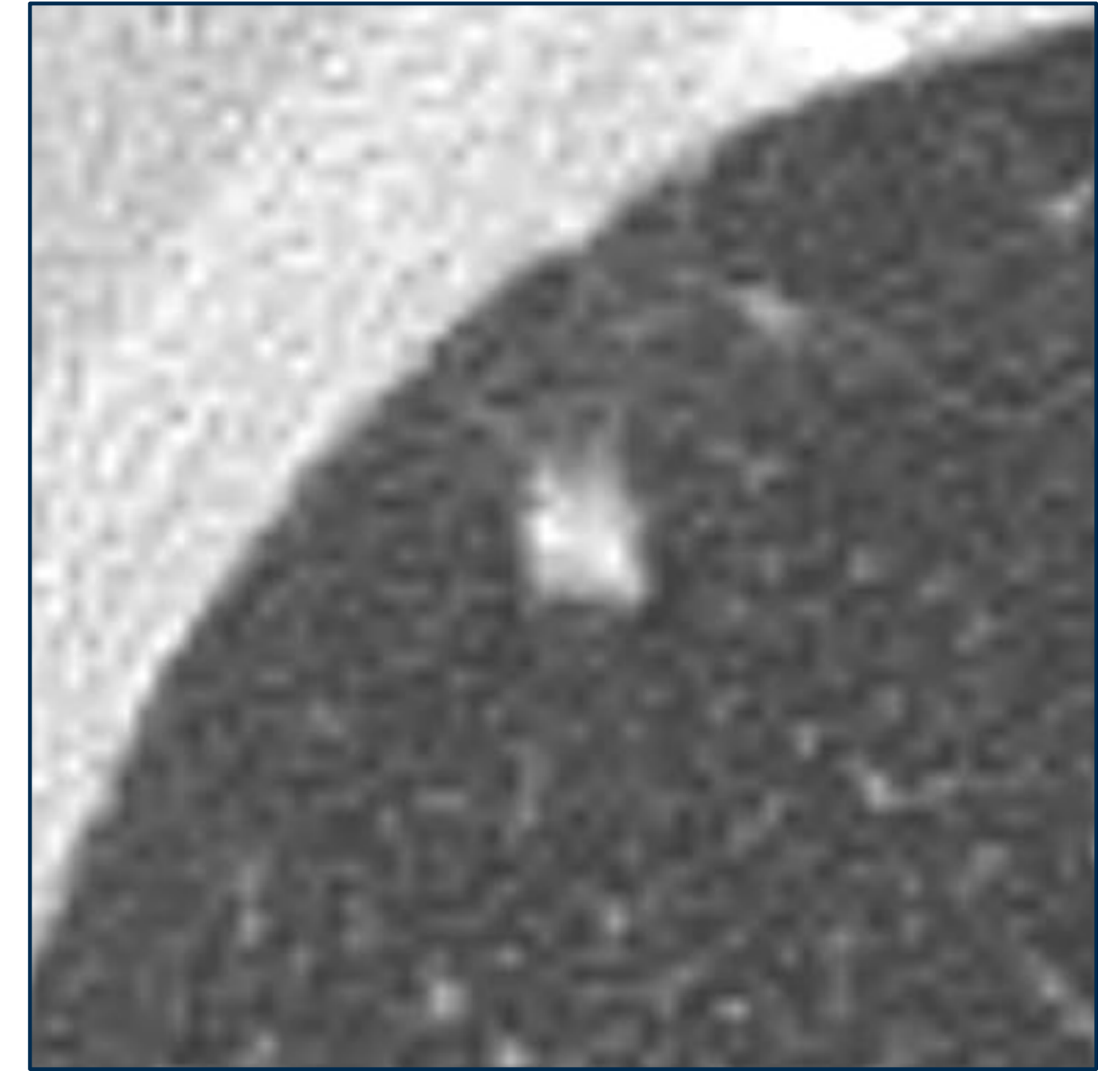
10mm



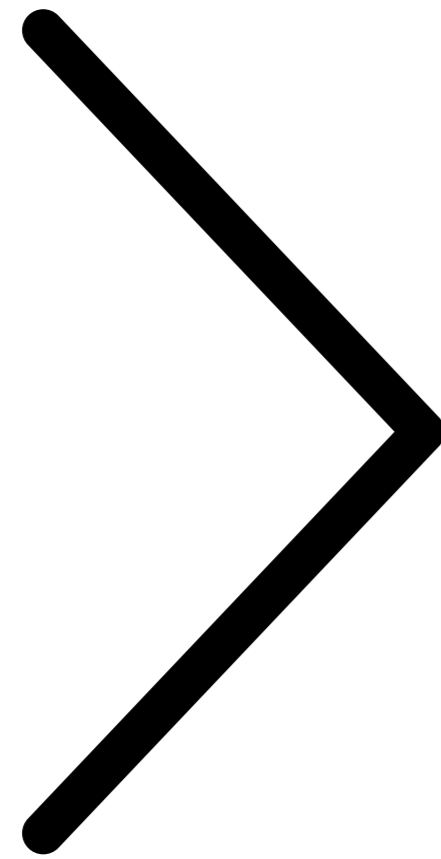
62yo M
8mm RML
LCP score: 2



48yo F
9mm RUL
LCP score: 5



47yo M
7mm RUL
LCP score: 3



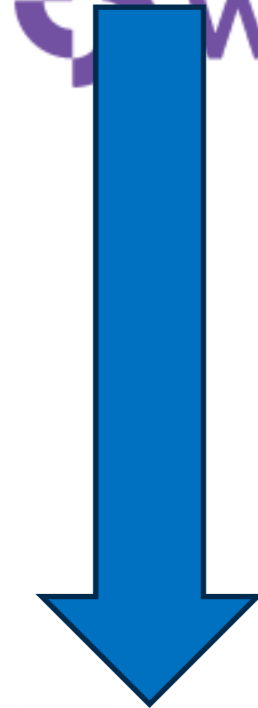
EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

MITIGATE RISK

PATHWAYS: NOPATIENT LEFT BEHIND

PATHWAY MANAGEMENT

WELLSTAR LCO: NO PATIENT LEFT BEHIND



Name	MRN	Gender	Age	Creation Date	Status	Sub-status	Defer/Discard Reason	Unreviewed report	Due date	Notes	Assignee	Managed By Facility	Tags	Actions
		Male	70		Diagnostic Follow-up	Order		0	03 Apr 2024	4		WellstarMF_PH	≥ 8 mm - < 15 mm CT Now	⋮

Name	MRN	Gender	Age	Creation Date	Status	Sub-status	Defer/Discard Reason	Unreviewed report	Due date	Notes	Assignee	Managed By Facility	Tags	Actions
		Male	70		Diagnostic Follow-up	Order		0	12 Oct 2024	2		WellstarMF_NF	< 6 mm Density 12 months	⋮

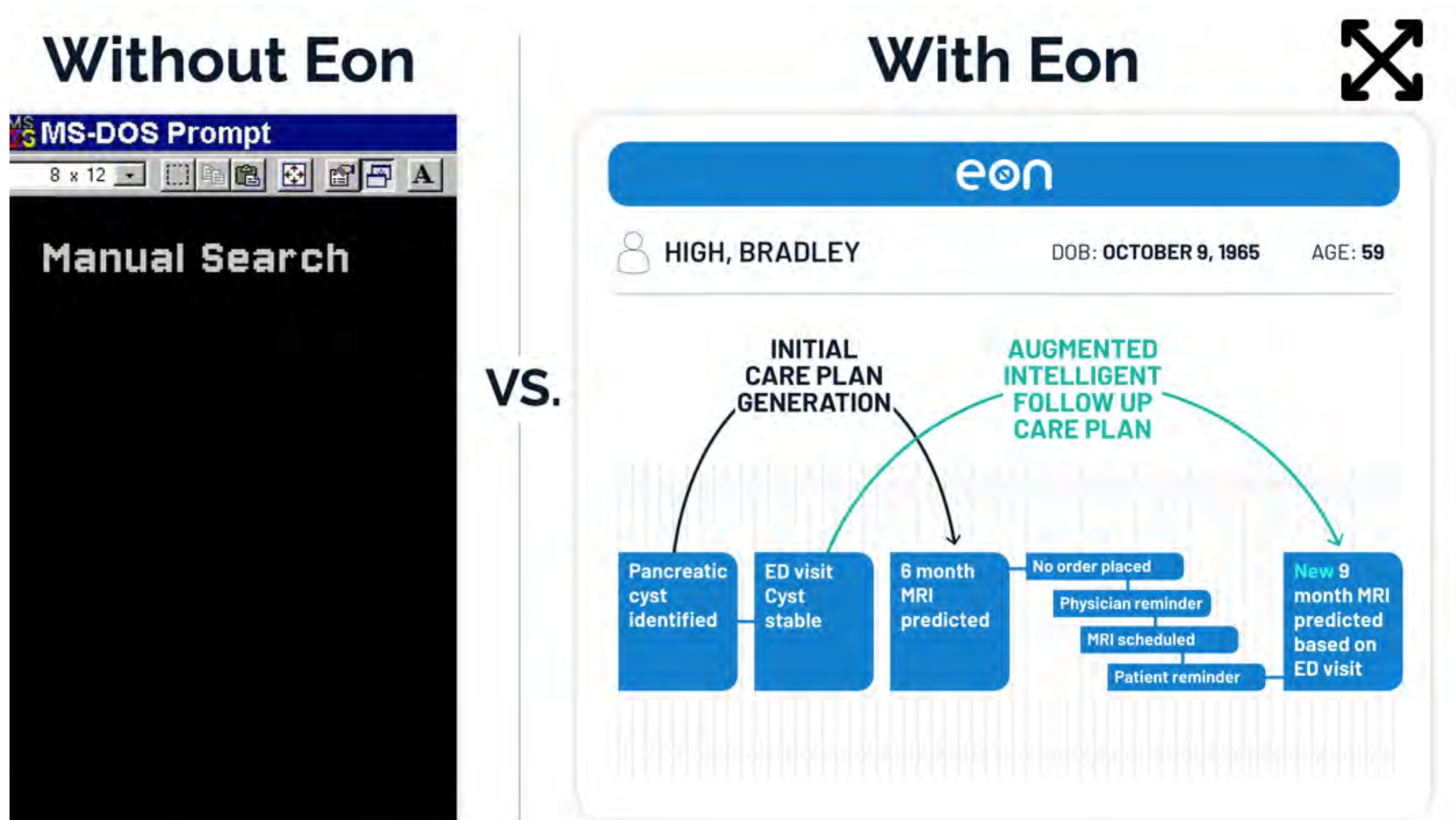
Manage Tags

Tags:

1 cm Send Letter Call Patient 3 months CT Now Biopsy Pending PFT/PET Pending ILN Clinic |

Cancel Confirm

PATHWAY MANAGEMENT

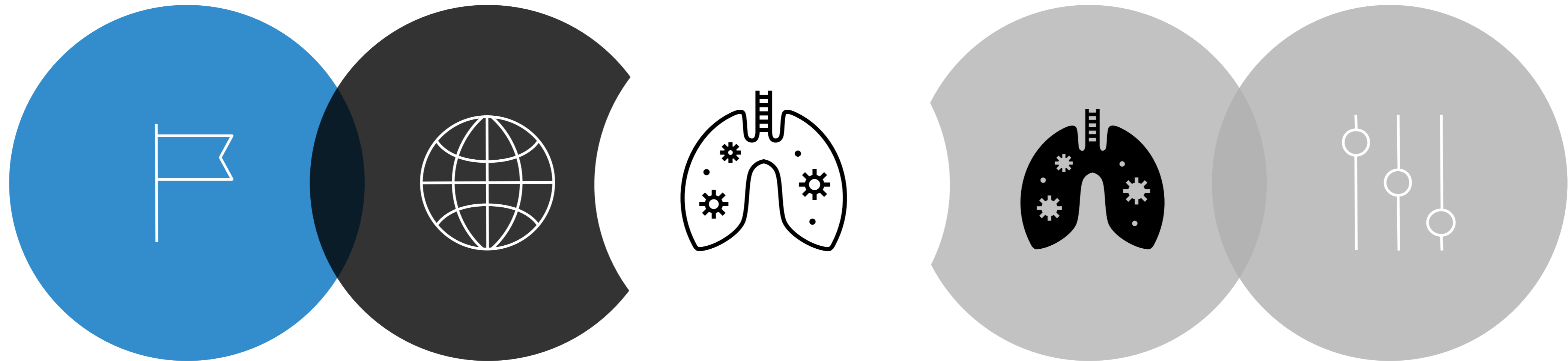


USUAL AND CUSTOMARY CARE **SPENDS 65 DAYS FROM ABNORMAL IMAGE TO DIAGNOSIS.**

AUTOMATED PATHWAYS SPEED THAT PROCESS.

LIMITATIONS OF CURRENT AI APPLICATIONS

IT'S US



INCOMPLETE DATA

SMOKING HISTORY,
RISK FACTOR DATA
FOR MODELS

INCONSISTENT TERMINOLOGY

10 RADIOLOGISTS =
TEN DIFFERENT
DESCRIPTIONS

INCONSISTENT NODULE DETECTION

INTER-READER
VARIABILITY

INCONSISTENT NODULE MEASUREMENT

GROSSLY INACCURATE
USE OF CURSOR

UNSTRUCTURED DATA

AI HAS A HARD TIME
PARSING OUT PHRASES
AND NUMBERS

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER.

**“EARLY DETECTION”
MEANS JUST THAT:**

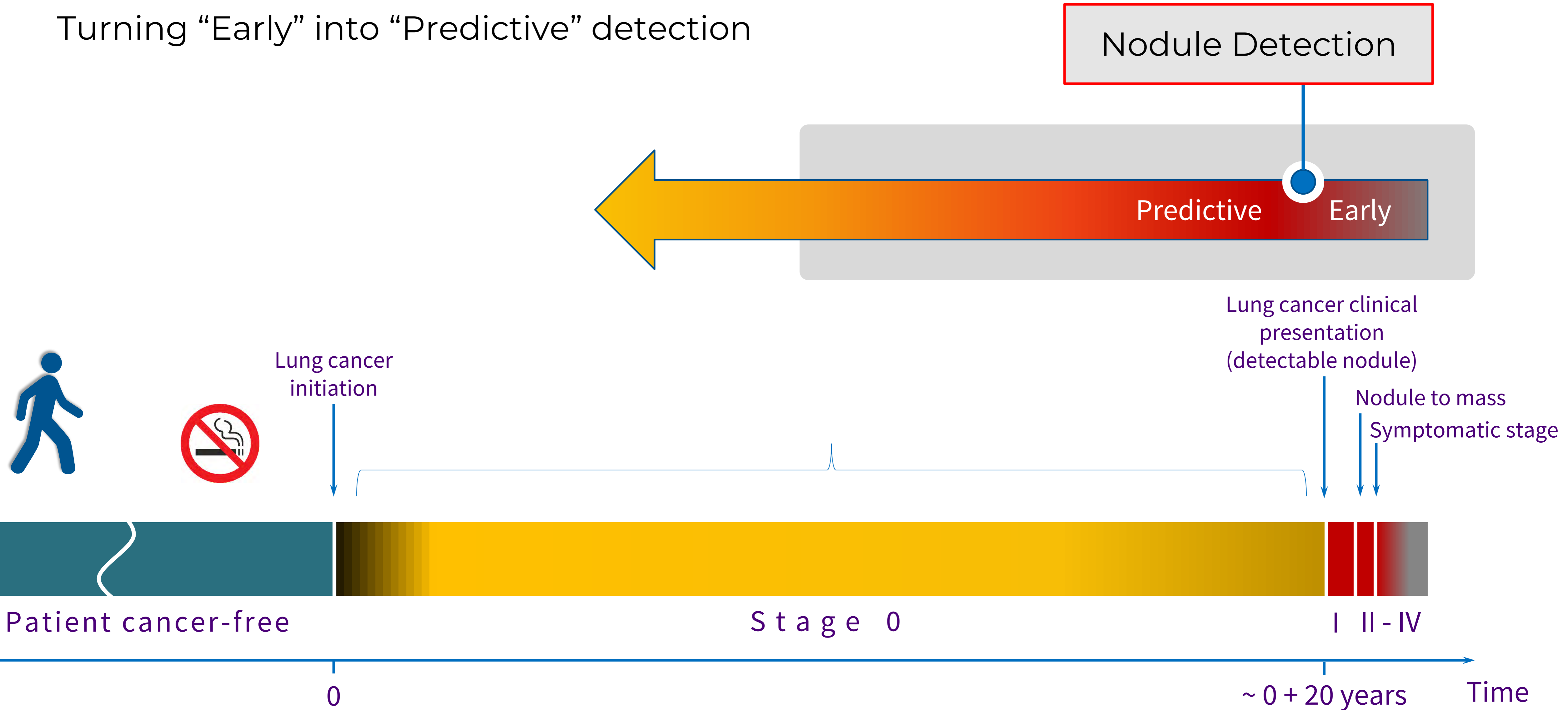
***YOU ALREADY HAVE A
CANCER AND WE NEED
TO FIND IT EARLY***

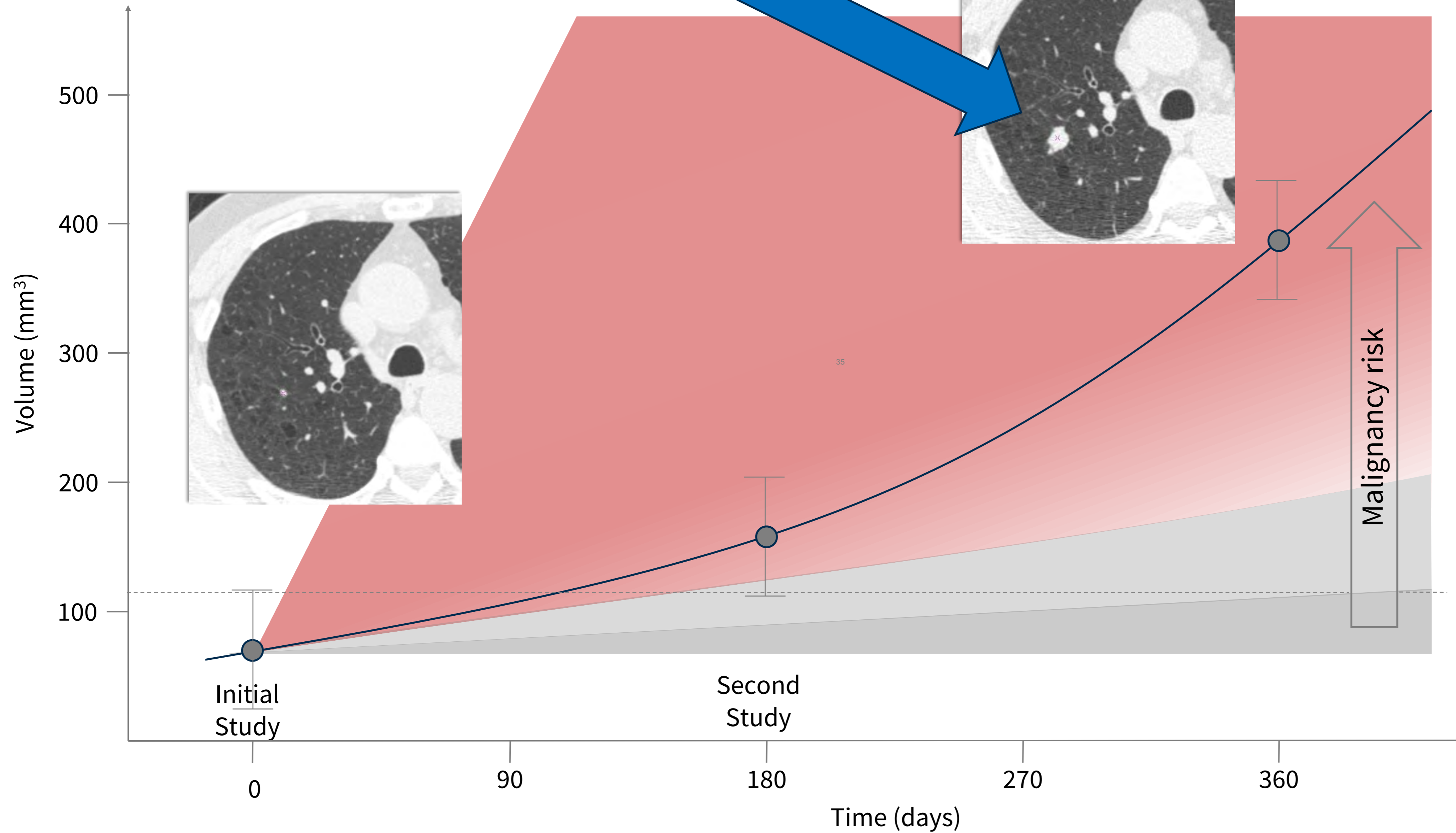
**CAN WE SHIFT IMAGING
FROM “REACTIVE”
TO “PREDICTIVE”**



Lung Cancer Timeline:

Turning "Early" into "Predictive" detection



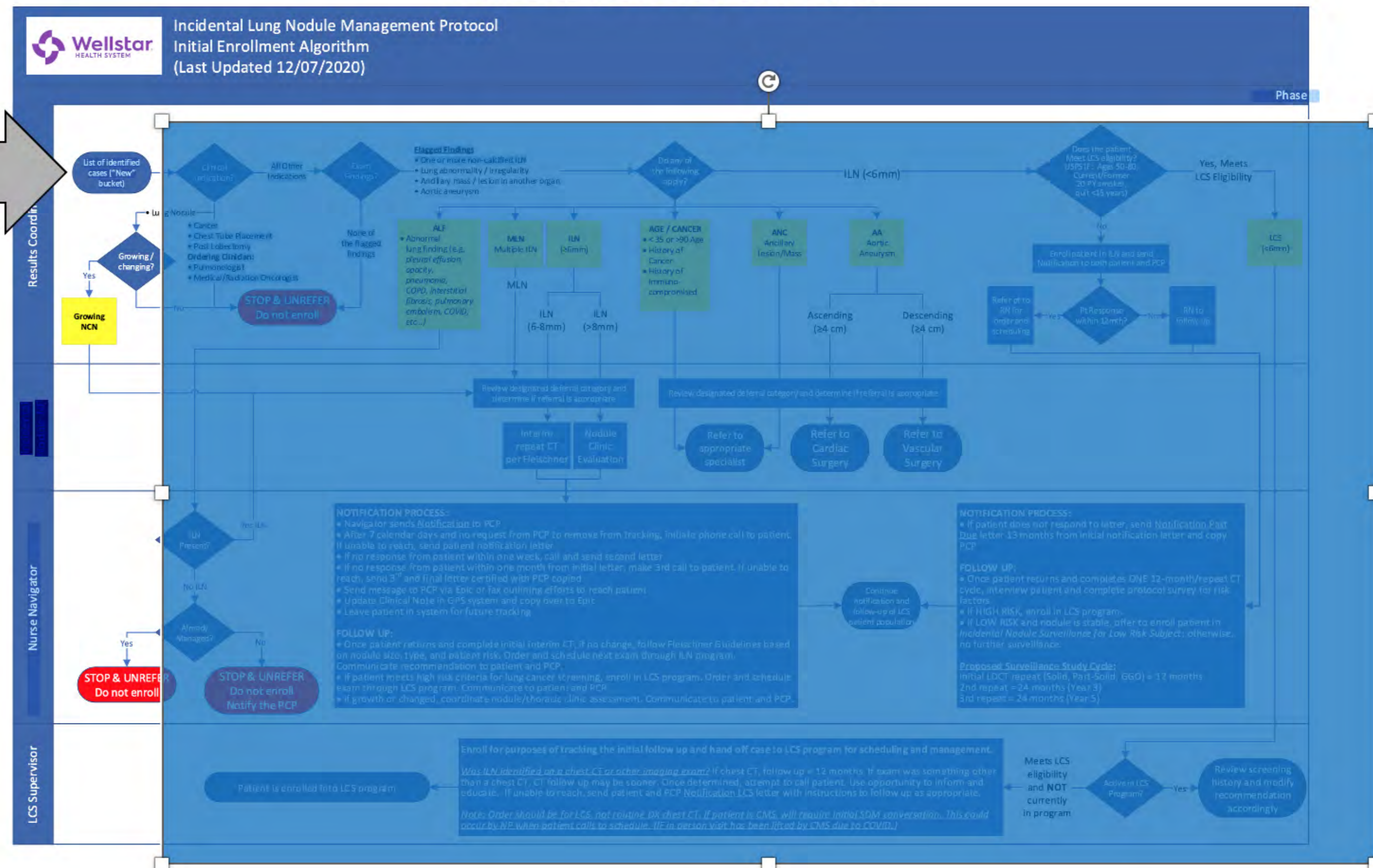


Depicted solid nodule growth pattern is an approximation based on suggestions in: "Growth patterns of screen-detected malignant pulmonary nodules: Accuracy of doubling-time models" A Creamer et al. European Respiratory Journal 2023; 62: Suppl. 67, OA3266 and is not necessarily indicative of the real clinical development

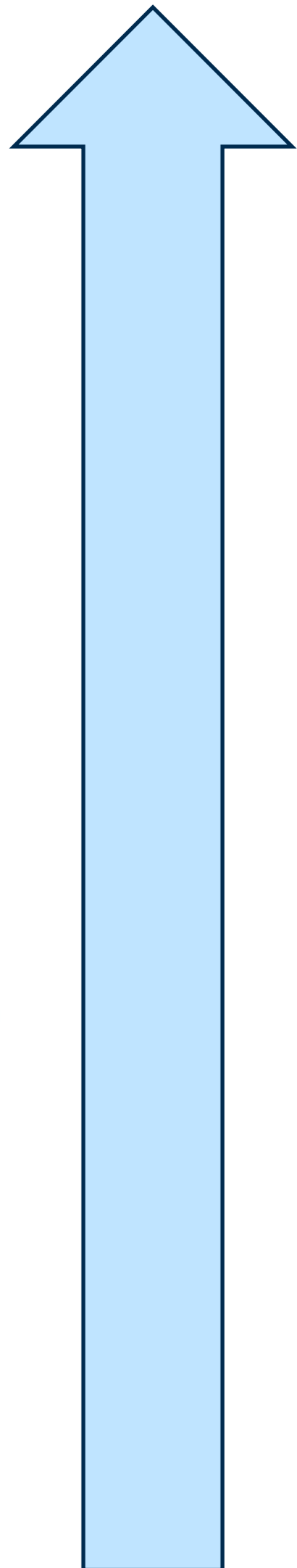
PREDICT

NO ONE DESERVES TO DIE OF LUNG CANCER

70%



WHAT IF ?



OPEN ACCESS | ORIGINAL REPORTS |  | January 12, 2023



Sybil: A Validated Deep Learning Model to Predict Future Lung Cancer Risk From a Single Low-Dose Chest Computed Tomography

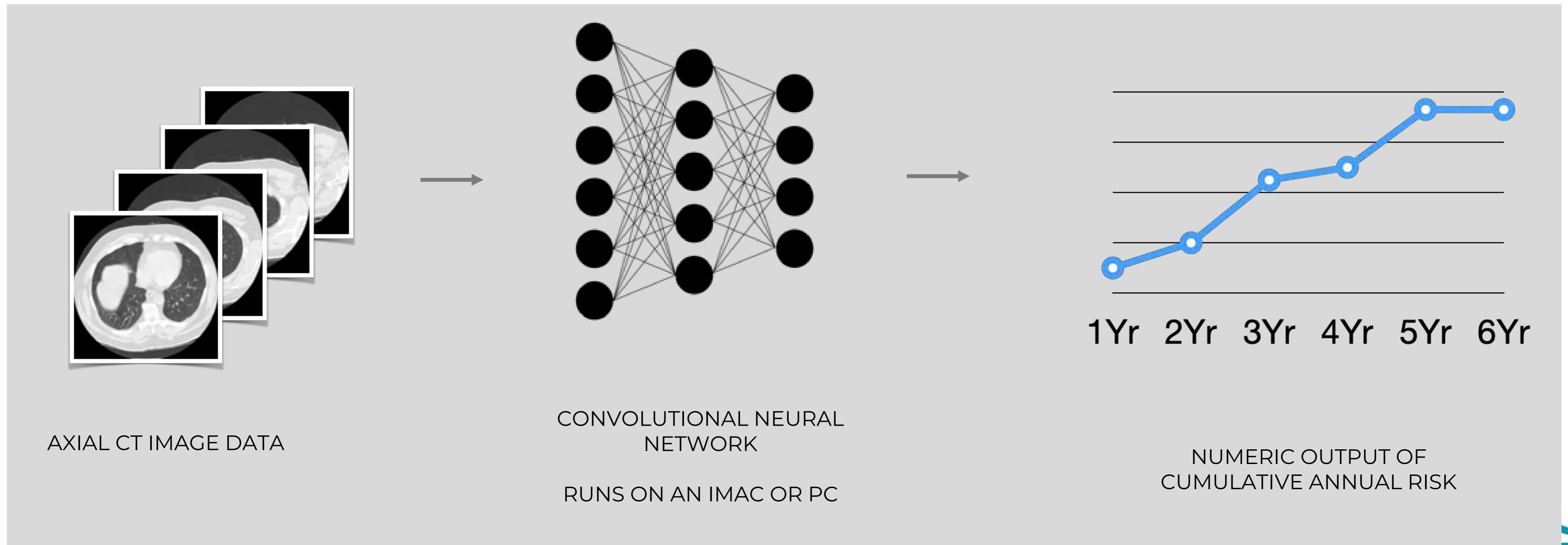
Authors: [Peter G. Mikhael, BSc](#) , [Jeremy Wohlwend, ME](#), [Adam Yala, PhD](#) , [Ludvig Karstens, MSc](#) , [Justin Xiang, ME](#), [Angelo K. Takigami, MD](#) , [Patrick P. Bourgoquin, MD](#) , ... [SHOW ALL ...](#), and [Regina Barzilay, PhD](#) | [AUTHORS INFO & AFFILIATIONS](#)

Publication: Journal of Clinical Oncology • [Volume 41, Number 12](#)
<https://doi.org/10.1200/JCO.22.01345>

NOT FDA APPROVED

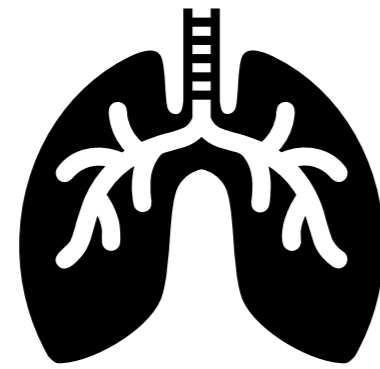
SYBIL

Deep Learning Algorithm trained on > 44K LDCT exams from 15K subjects in **NLST trial**



SYBIL PREDICTS LUNG CANCER FROM A SINGLE LDCT

Mikhael, *J Clin Oncol* 2023

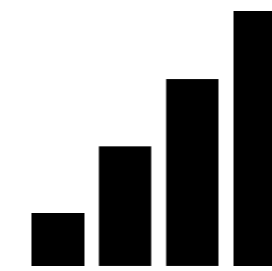


SINGLE CT

LOW DOSE CT OF THE CHEST

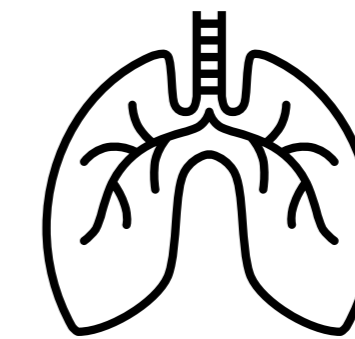
NO CLINICAL INPUT
REQUIRED

NO NODULE REQUIRED



IMMEDIATELY REPORTS 1 – 6 YEAR RISK

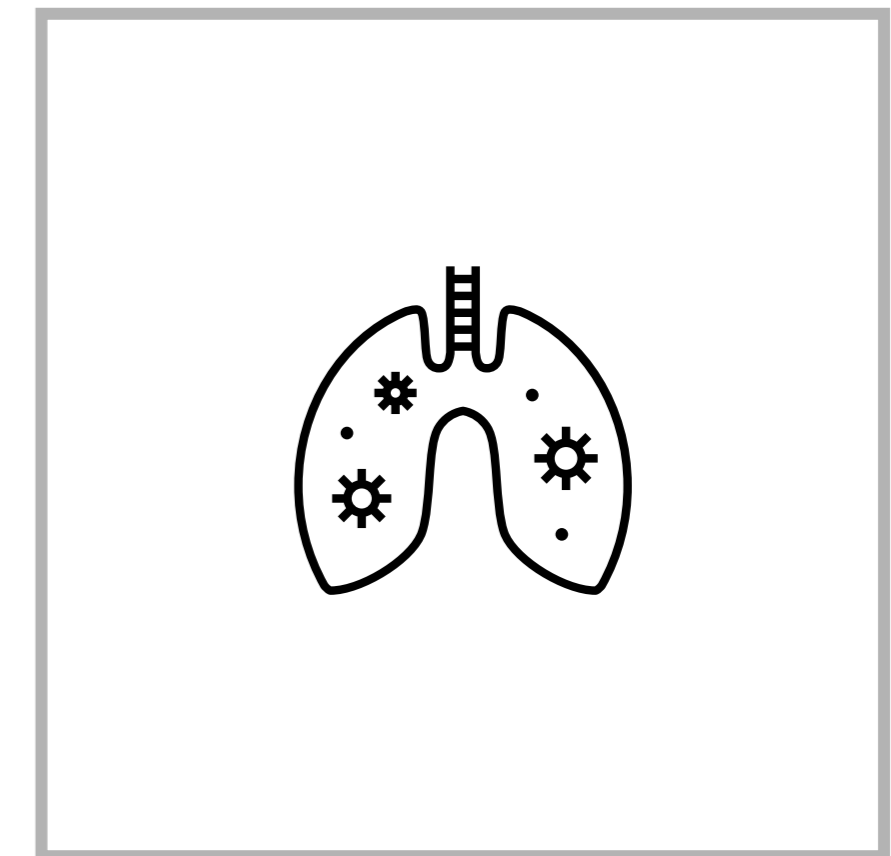
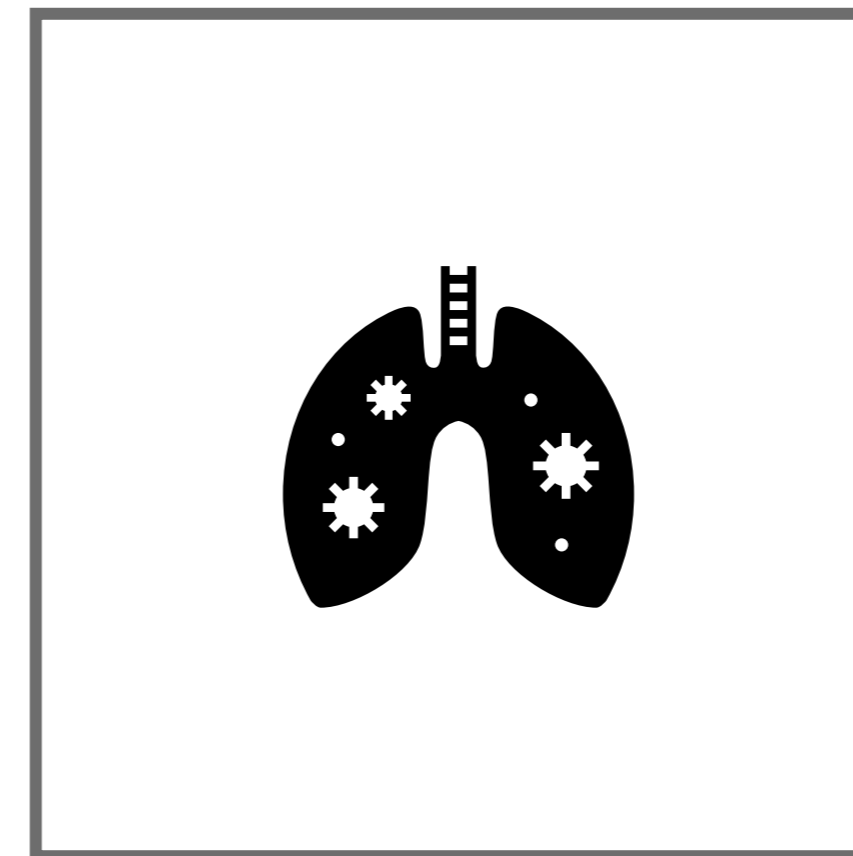
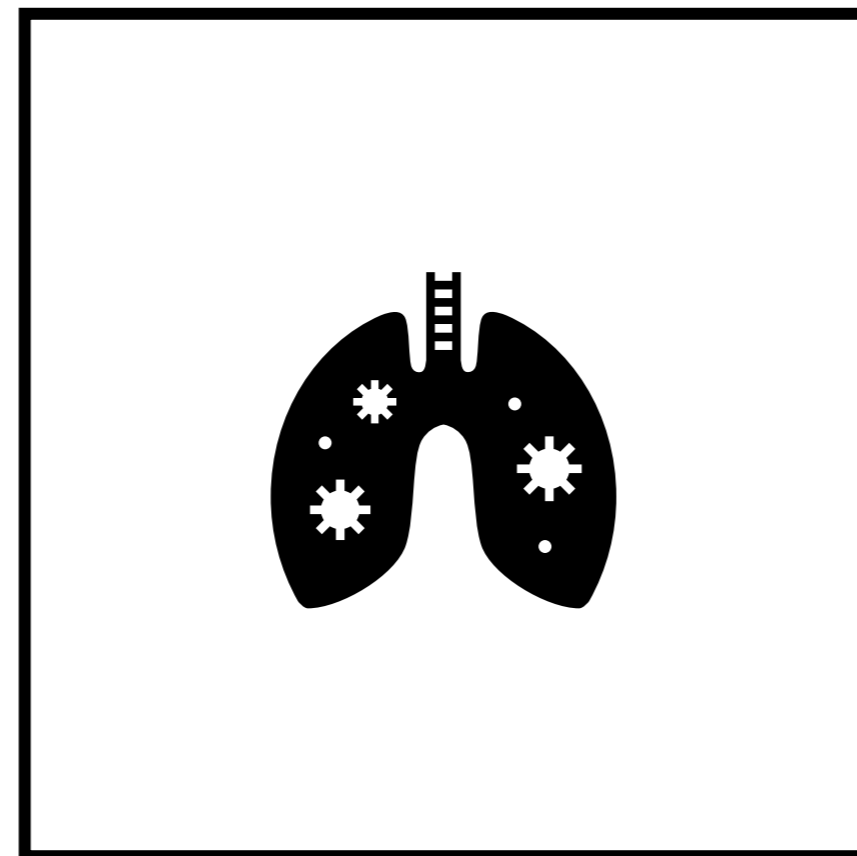
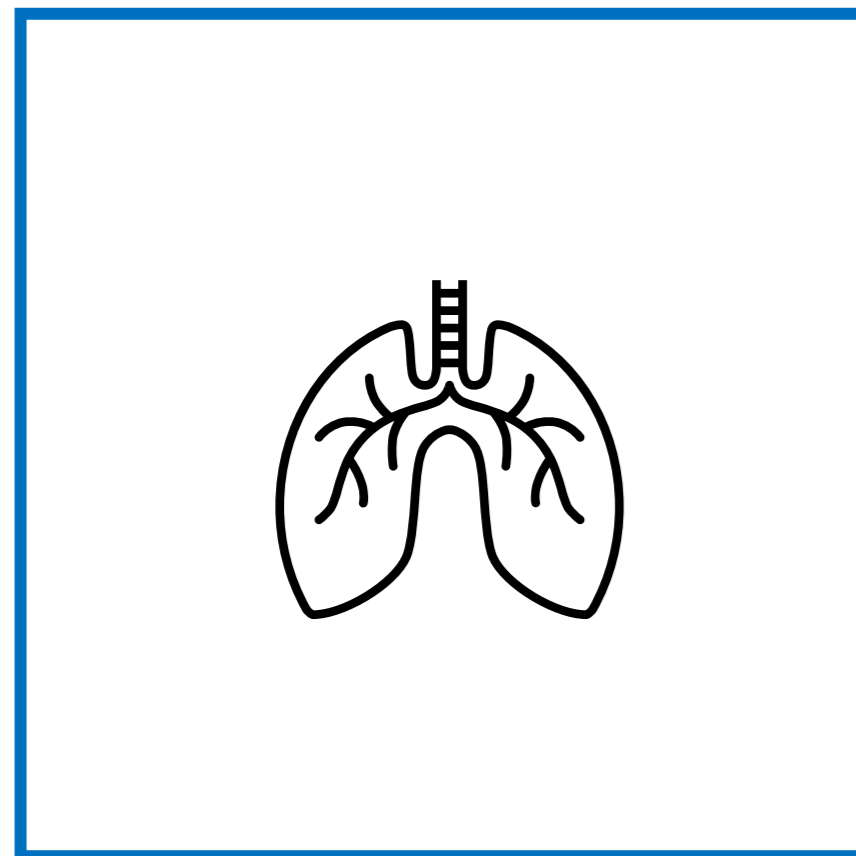
NO EXTERNAL
TRANSFER OF
IMAGES OR DATA



NO NODULE OR ANNOTATION REQUIRED

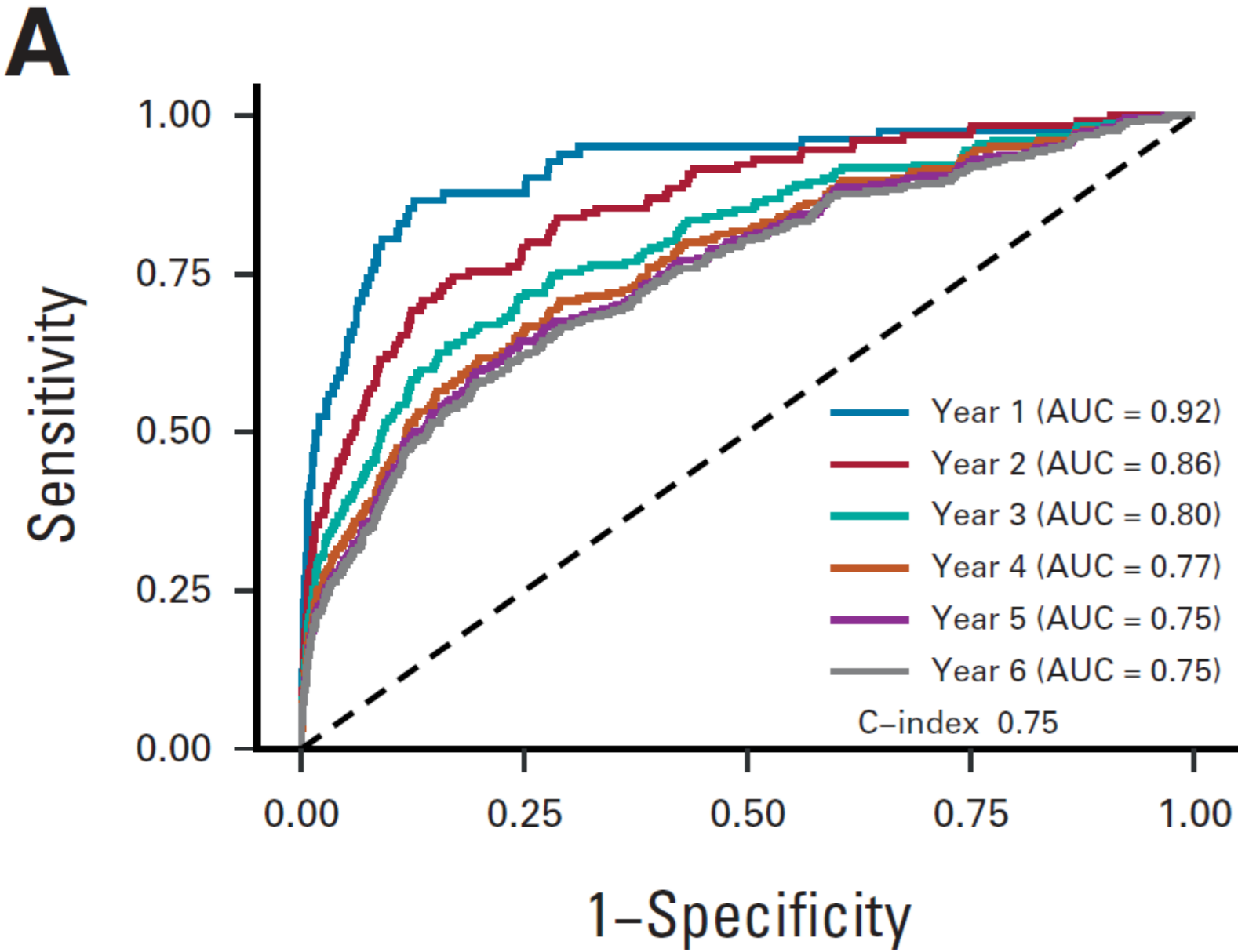
SYBIL “READS” THE LDCT
AND PREDICTS FUTURE
LUNG CANCER

SYBIL SIMPLY READS THE LDCT

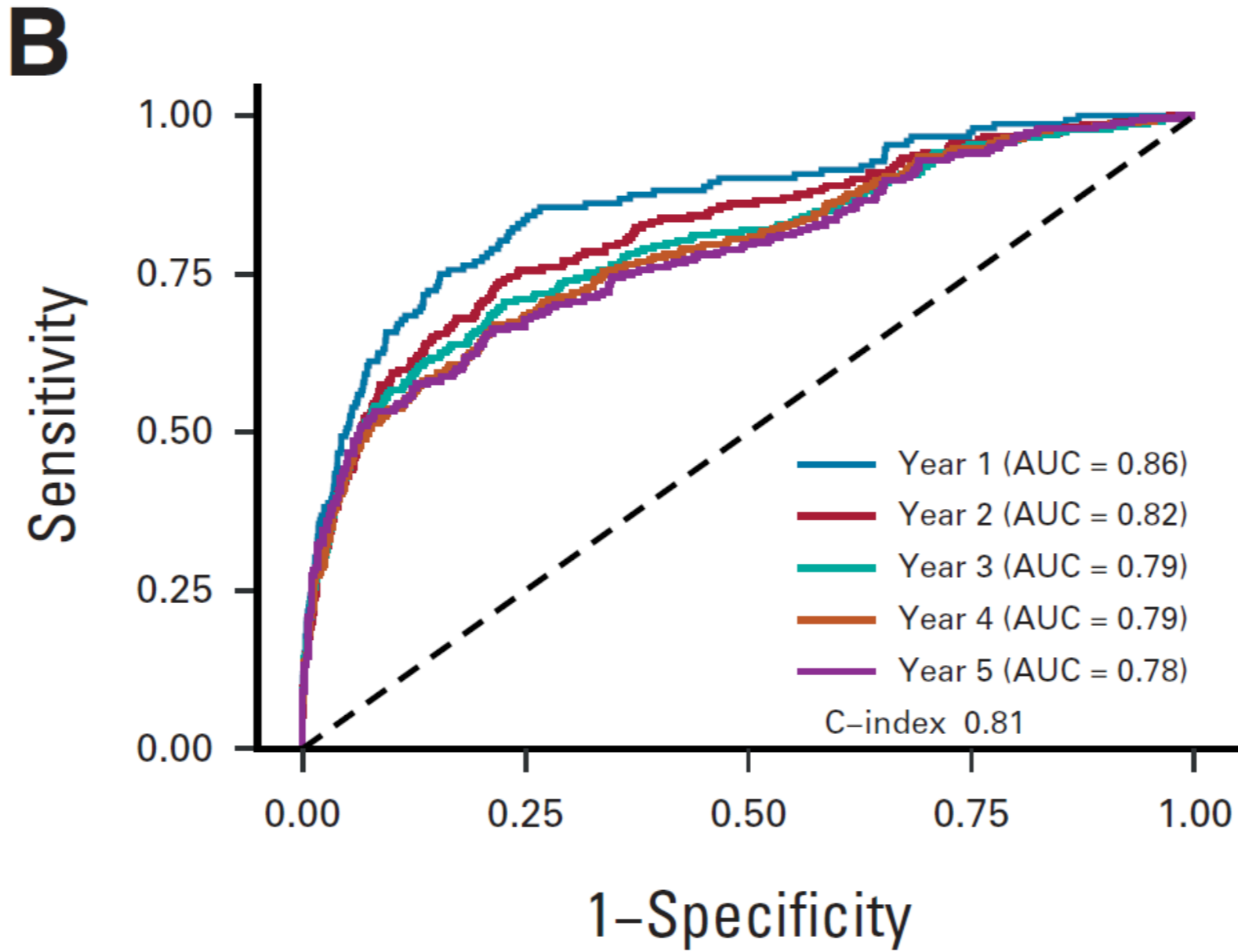


**“EARLY DETECTION IS GOOD.
CANCER PREDICTION IS BETTER.”**

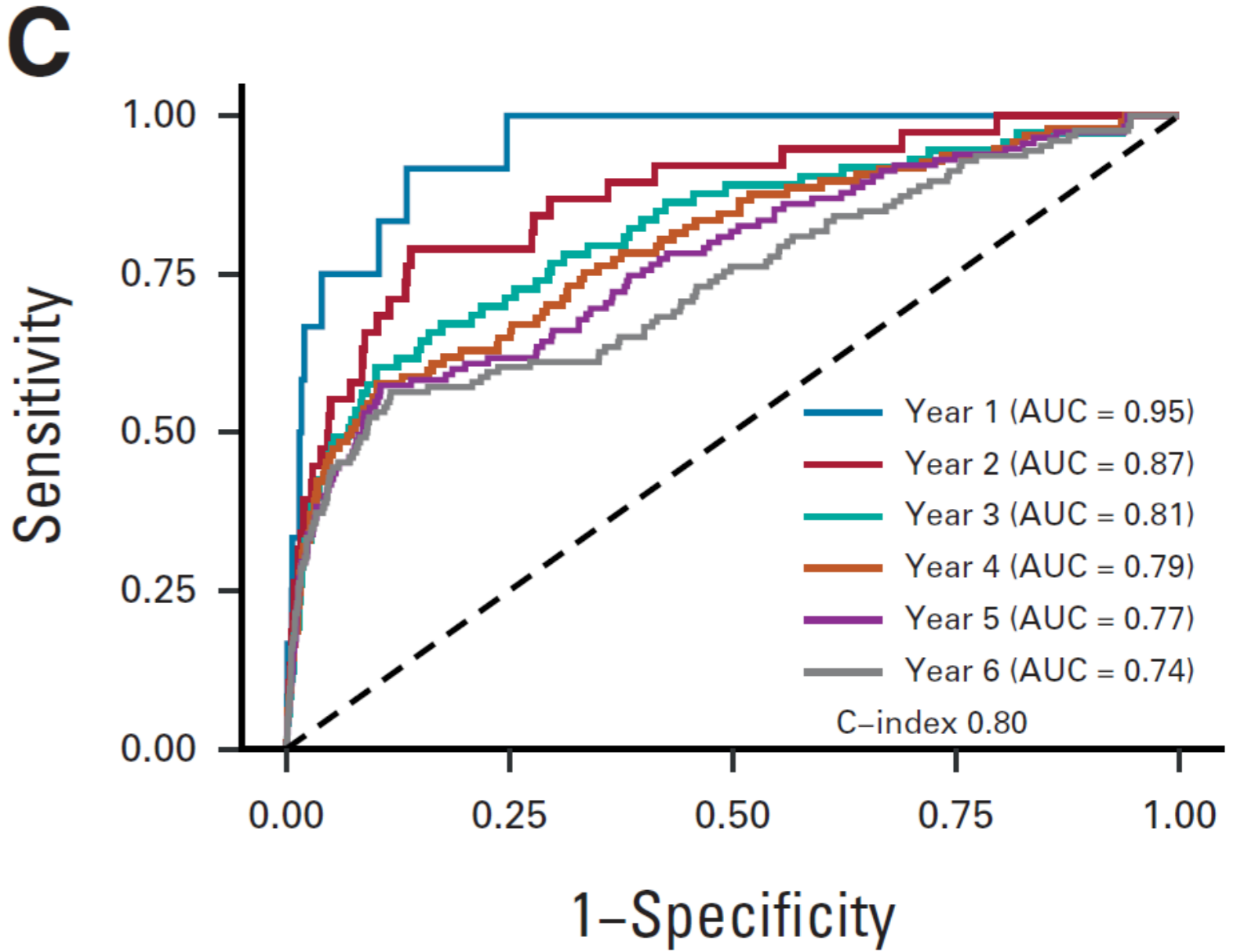
Sybil Performance



NLST, n=6,282



MGH, n=8,862



CGMH, n= 12,280



KEY ADVANTAGES



No image annotation

No clinical data needed

Instant readouts

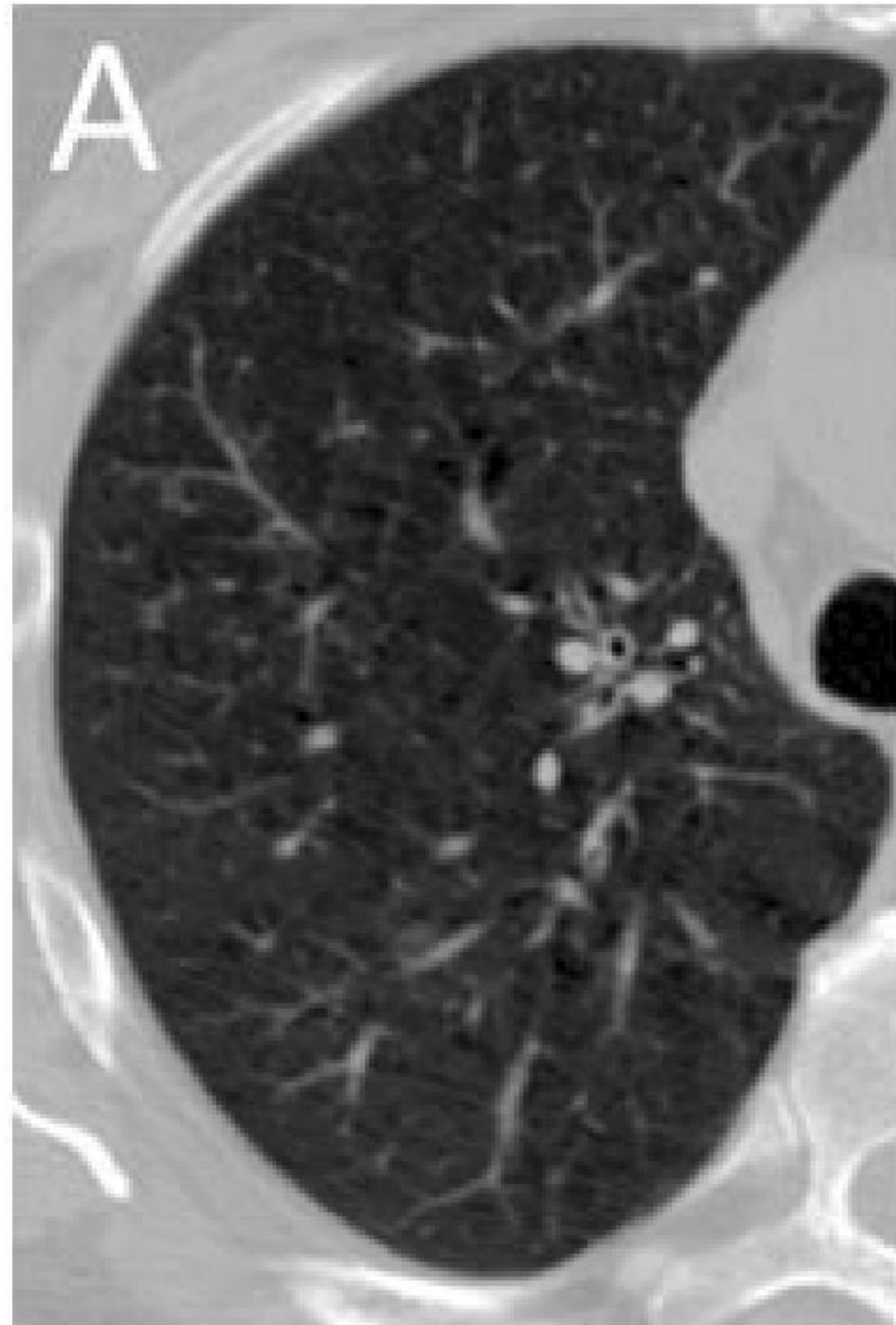
Cumulative multi-year future risk stratification from a single LDCT

Most Common Question



WHAT IS SYBIL SEEING?

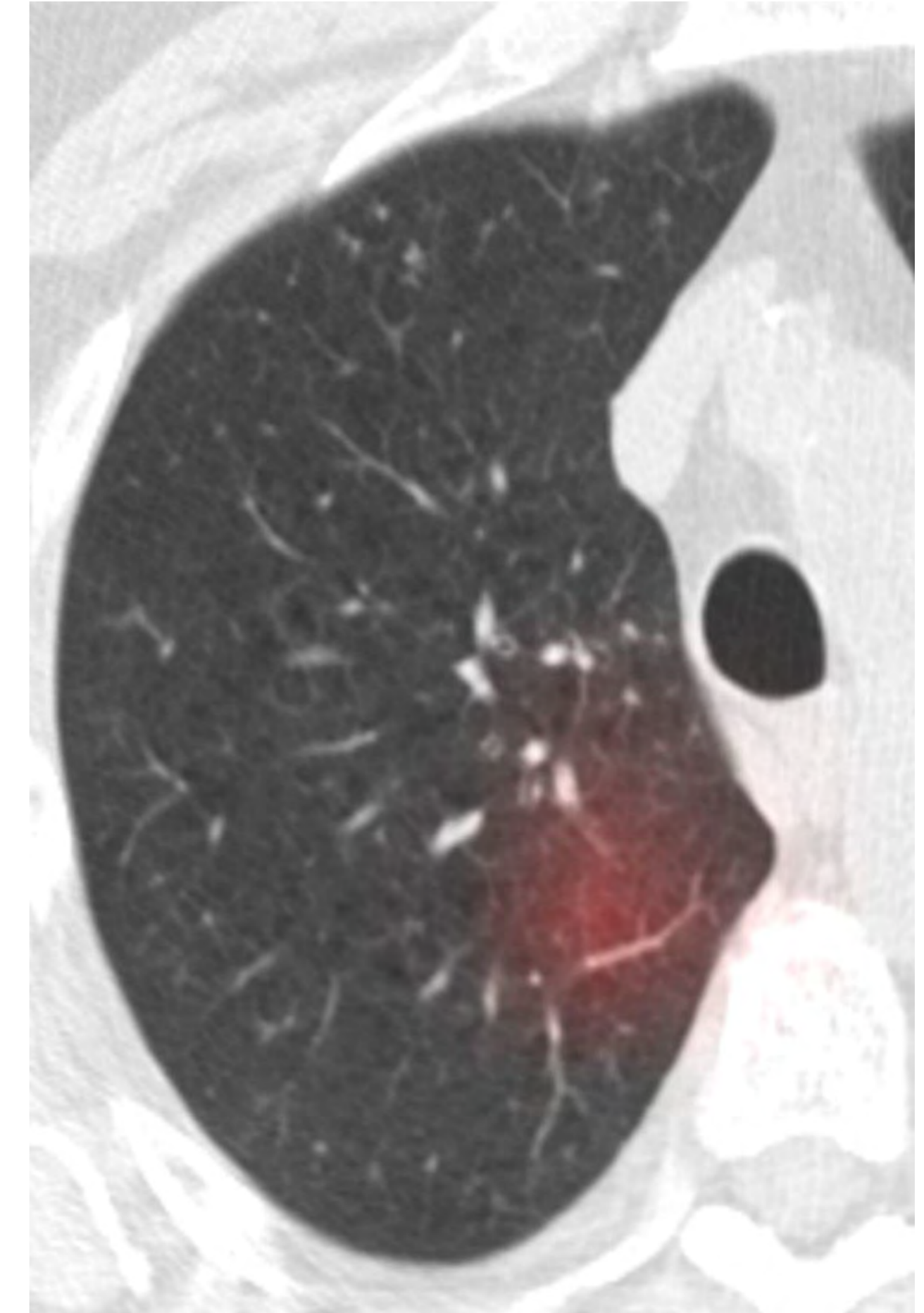
What is Sybil Seeing?



69-year-old male with
99 pack-year smoking hx

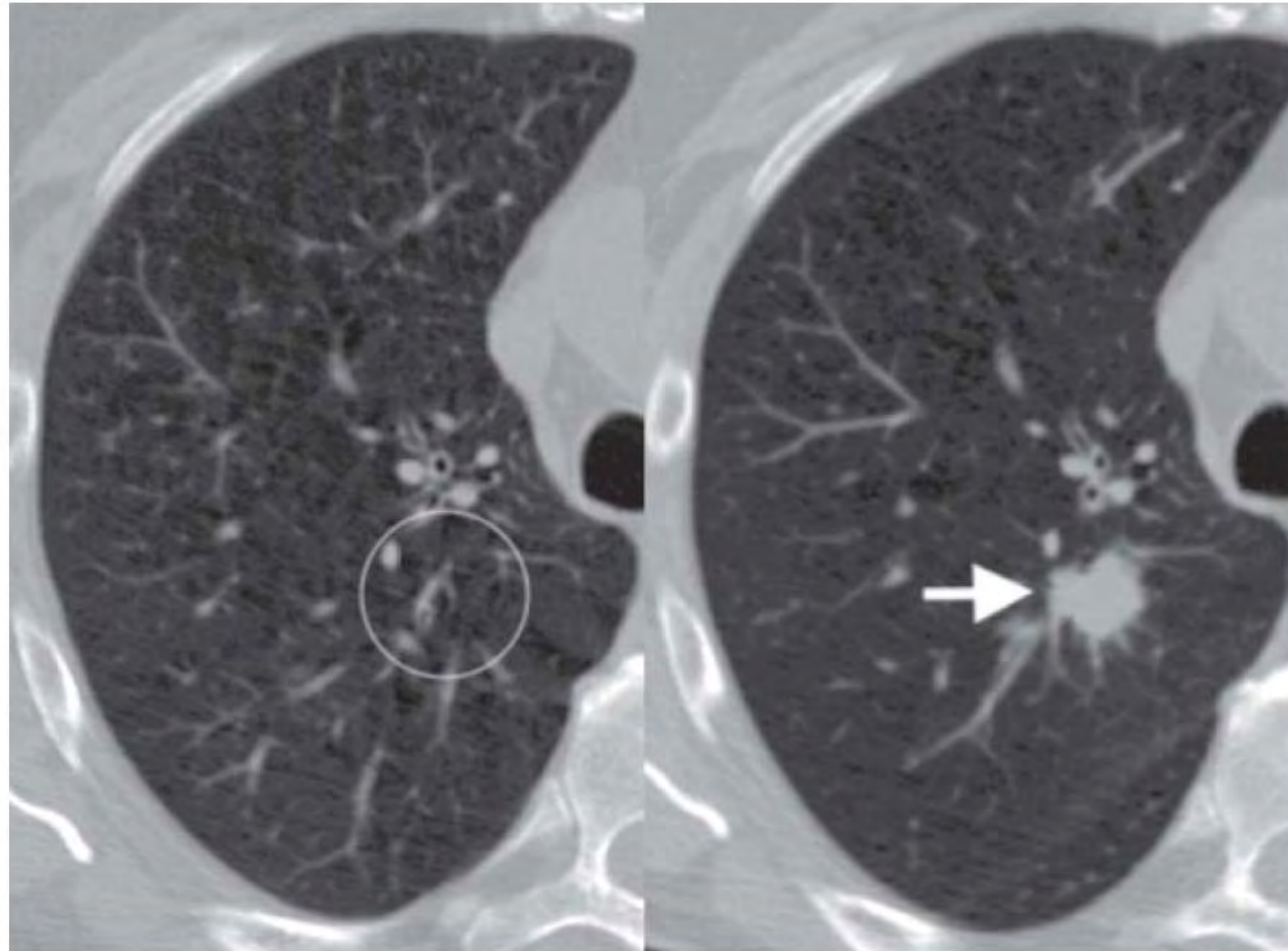
Baseline scan read as “negative
screen, minor abnormalities not
suspicious for lung cancer”

Sybil placed scan in **75% risk
percentile** (6-year risk)



Sybil “attention map”
(note this is *not* a PET scan)

What is Sybil Seeing?



Mikhael, *J Clin Oncol* 2023

One year followup

2.2 cm poorly-differentiated squamous cell lung cancer

Surgical resection (pT1cN0)

SYBIL READS THE RAW LDCT DATA,
NOT THE VISUAL ALGORITHM

HUMANS SEE THE VISUAL ALGORITHM



EARLY DETECTION IS GOOD. CANCER PREDICTION IS BETTER.

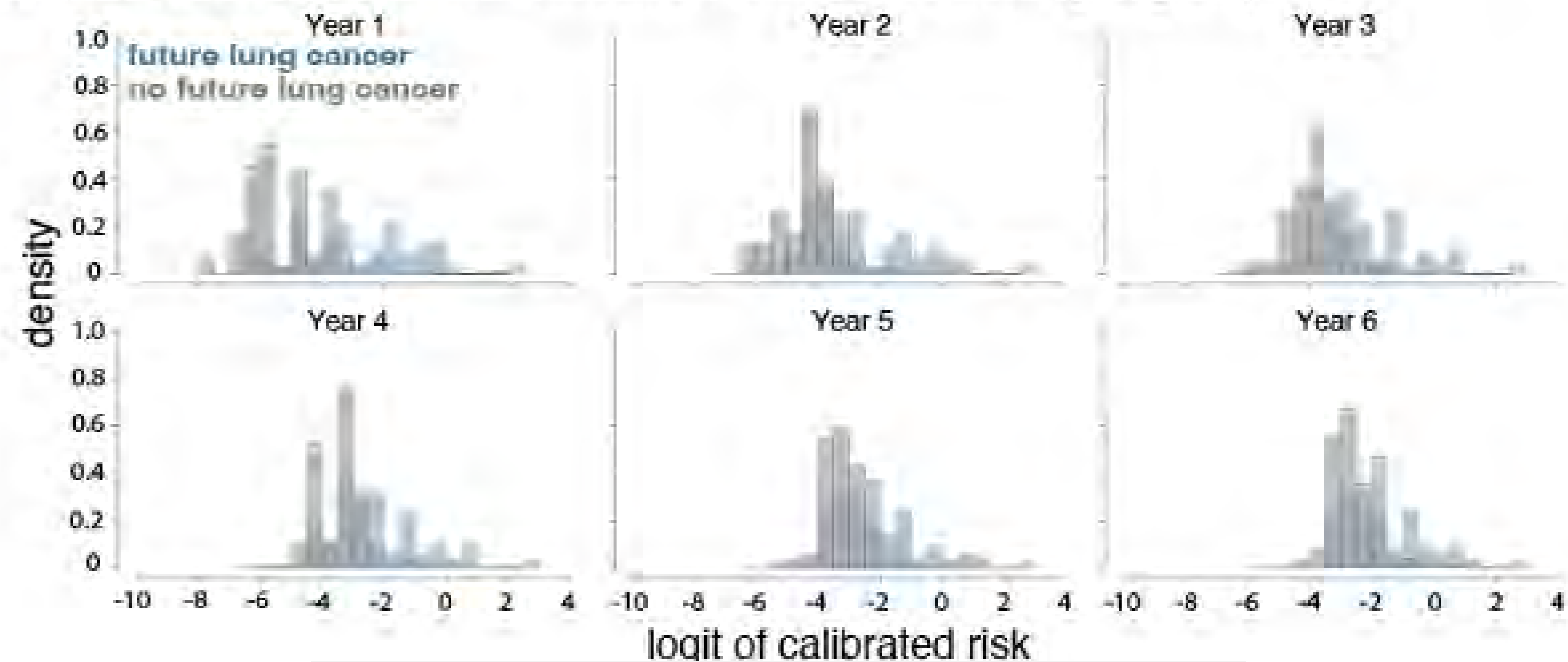
CLINICAL UTILITY

70% ACCURATE IN SELECTING PATIENTS WHO WILL DEVELOP LUNG CANCER

A LOW SYBIL SCORE WILL NOT PREVENT FURTHER FOLLOW UP

A HIGH SYBIL SCORE WILL PROMOTE MORE CAREFUL FOLLOW UP

D. Mass General external validation set (US screening eligible)



EARLY DETECTION IS GOOD. CANCER PREDICTION IS BETTER.

SYBIL CONSORTIUM

VALIDATION STUDIES:

DIVERSE POPULATION OF SUBJECTS

HIGH VOLUMES

STANDARDIZED DATA SET

PROSPECTIVE VALIDATION OF DLA IN
INCIDENTALS

RETROSPECTIVE VALIDATION OF DLA



EARLY DETECTION IS GOOD. CANCER PREDICTION IS BETTER.

LUNG CANCER PREDICTION

SYBIL

READS LDCT
SIX YEAR RISK

LUMAS

READS LDCT
ONE YEAR RISK

XGBOOST

EMR DATA MINING
ONE YEAR RISK

nature
medicine

LETTERS

<https://doi.org/10.1038/s41591-019-0447-x>

End-to-end lung cancer screening with three-dimensional deep learning on low-dose chest computed tomography

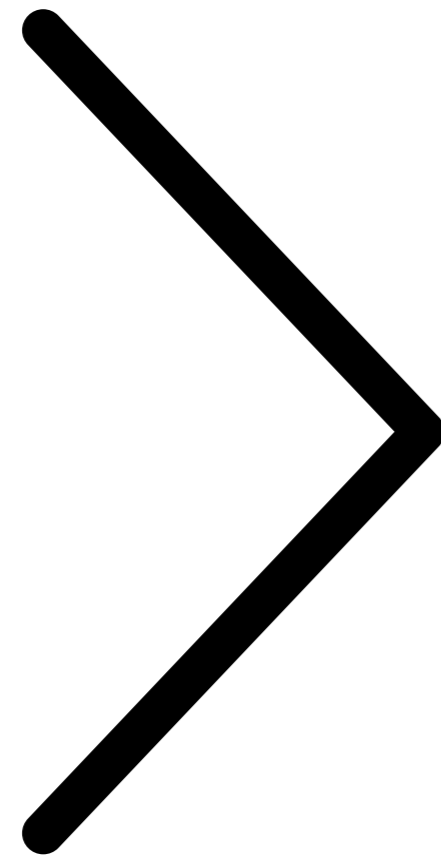
Diego Ardila^{1,5}, Atilla P. Kiraly^{1,5}, Sujeeth Bharadwaj^{1,5}, Bokyung Choi^{1,5}, Joshua J. Reicher², Lily Peng¹, Daniel Tse^{1*}, Mozziyar Etemadi³, Wenxing Ye¹, Greg Corrado¹, David P. Naidich⁴ and Shravya Shetty¹

With an estimated 160,000 deaths in 2018, lung cancer is the most common cause of cancer death in the United States¹. Lung cancer screening using low-dose computed tomography has been shown to reduce mortality by 20–43% and is now included in US screening guidelines^{1–6}. Existing challenges include inter-grader variability and high false-positive and false-negative rates^{7–10}. We propose a deep learning algorithm that uses a patient's current and prior computed tomography volumes to predict the risk of lung cancer. Our model achieves a state-of-the-art performance (94.4% area under the curve) on 6,716 National Lung Cancer Screening Trial cases, and performs similarly on an independent clinical validation set of 1,139 cases. We conducted two reader studies. When prior computed tomography imaging was not available, our model outperformed all six radiologists with absolute reductions of 11% in false positives and 5% in false negatives. Where prior computed tomography imaging was available, the model performance was on-par with the same radiologists. This creates an opportunity to optimize the screening process via computer assistance and automation. While the vast majority of patients remain unscreened, we show the potential for deep learning models to increase the accuracy, consistency and adoption of lung cancer screening worldwide.

limitations suggest opportunities for more sophisticated systems to improve performance and inter-reader consistency^{18,19}. Deep learning approaches offer the exciting potential to automate more complex image analysis, detect subtle holistic imaging findings and unify methodologies for image evaluation²⁰.

A variety of software devices have been approved by the Food and Drug Administration (FDA) with the goal of addressing workflow efficiency and performance through augmented detection of lung nodules on lung computed tomography (CT)²¹. Clinical research has primarily focused on either nodule detection or diagnostic support for lesions manually selected by imaging experts^{22–27}. Nodule detection systems were engineered with the goal of improving radiologist sensitivity in identifying nodules while minimizing costs to specificity, thereby falling into the category of computer-aided detection (CADe)²⁸. This approach highlights small nodules, leaving malignancy risk evaluation and clinical decision making to the clinician. Diagnostic support for pre-identified lesions is included in computer-aided diagnosis (CADx) platforms, which are primarily aimed at improving specificity. CADx has gained greater interest and even first regulatory approvals in other areas of radiology, though not in lung cancer at the time of manuscript preparation²⁹.

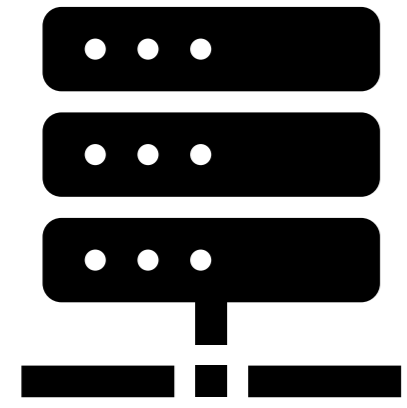
To move beyond the limitations of prior CADe and CADx



EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER

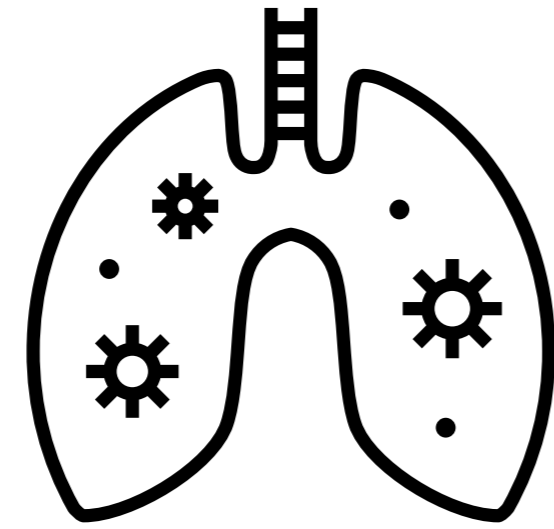
FUTURE

FDA APPROVED ARTIFICIAL INTELLIGENCE



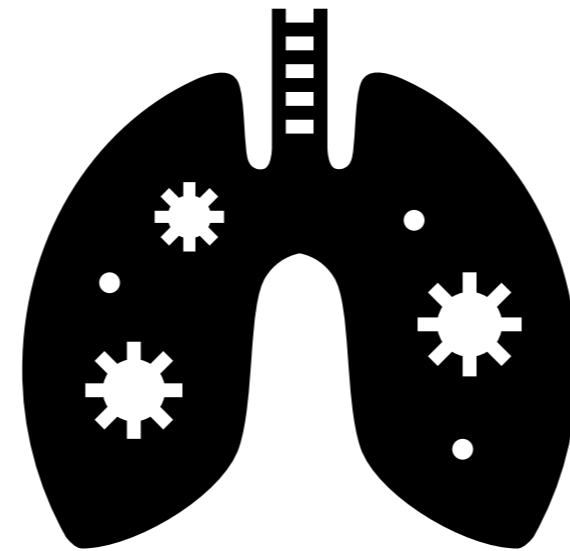
692

AI ENABLED
MEDICAL DEVICES
IN HEALTHCARE
(SAAD)



77%

RADIOLOGY



FEW

LUNG CANCER

AZRA AI

EON

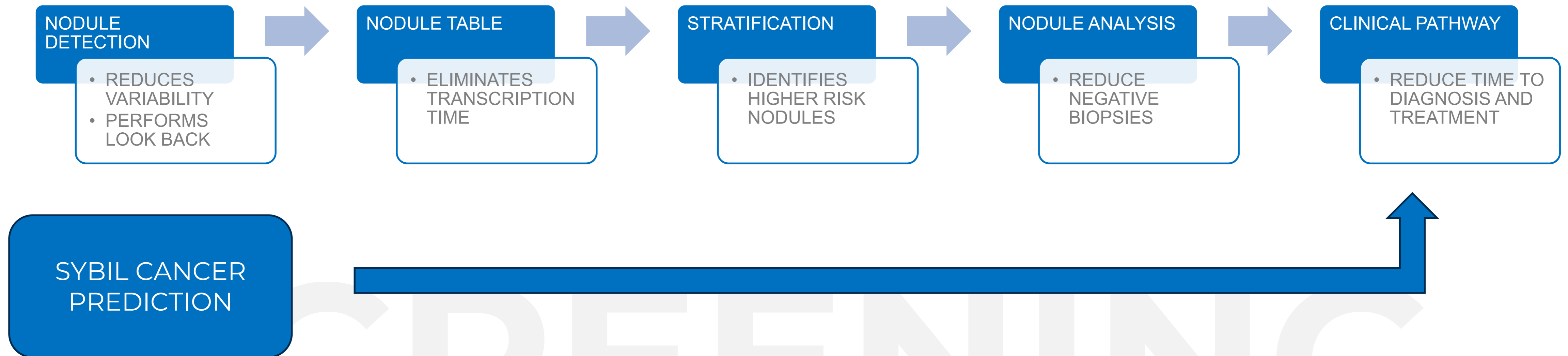
LUNG CANCER ORCHESTRATOR

RIVERAIN

OPTELLUM

MY DREAM TEAM OF FEATURES

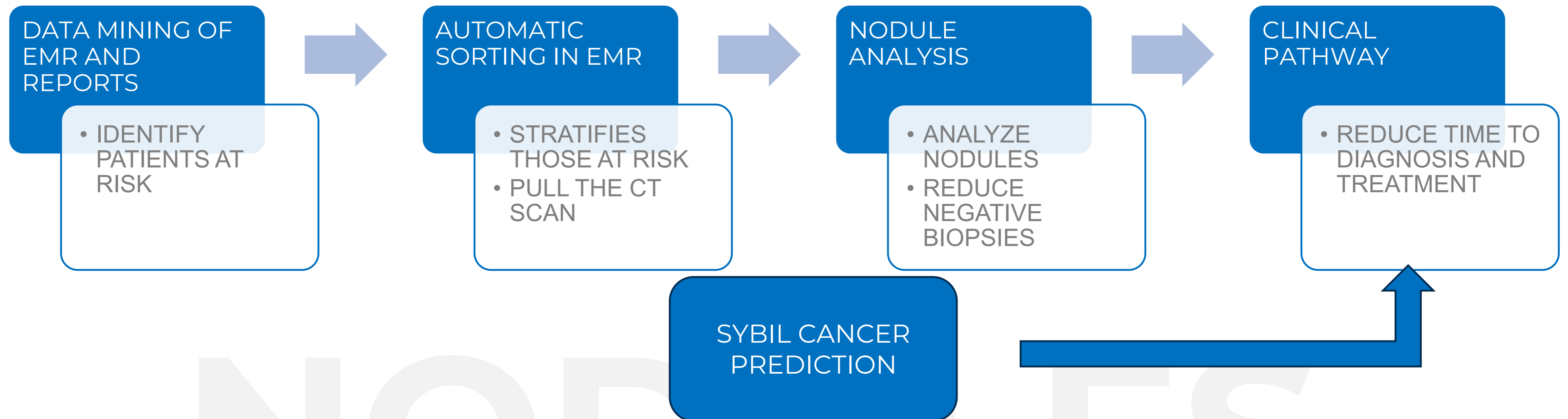
LUNG



SCREENING

MY DREAM TEAM OF FEATURES

INCIDENTAL



NODULES

AI WILL
DISRUPT OUR
CURRENT
MANUAL
PROCESSES

IMPROVE
EFFICIENCY
AND QUALITY

IDENTIFY RISK



STRATIFY RISK



ANALYZE RISK



MITIGATE RISK

PREDICT

IDENTIFY RISK

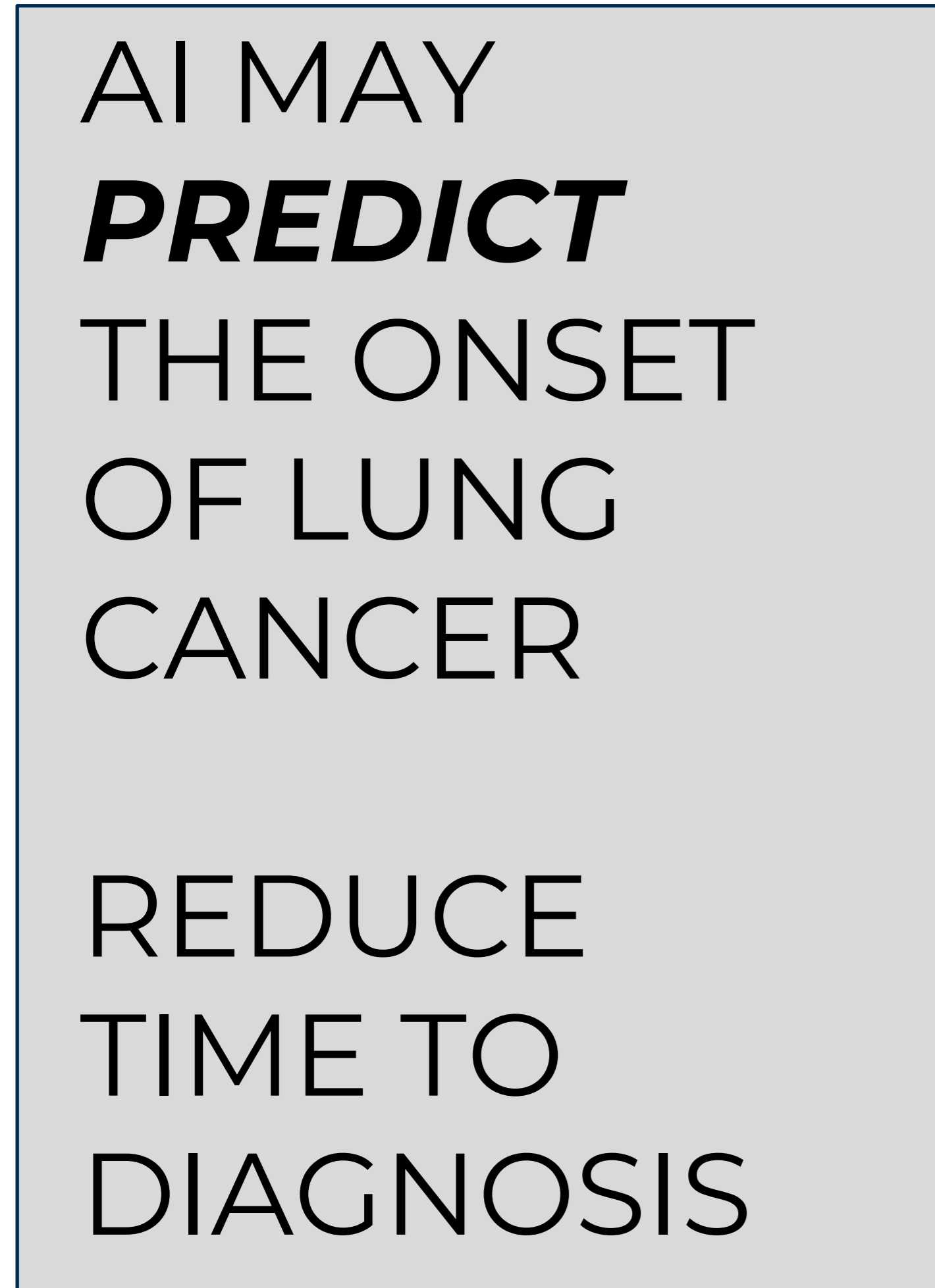
STRATIFY RISK

ANALYZE RISK

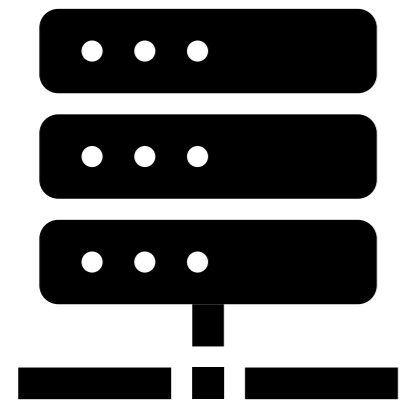
MITIGATE RISK

AI MAY
PREDICT
THE ONSET
OF LUNG
CANCER

REDUCE
TIME TO
DIAGNOSIS

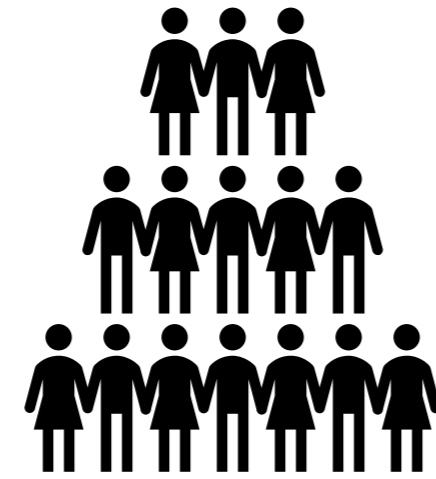


THE FUTURE for AI in Lung Cancer



Data Management

- **Convert Unstructured data to Structured**
- **Data Mining**



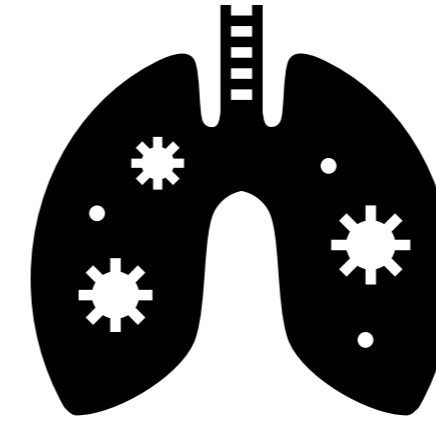
Population Health

- Refine Population Risk assessment
- Refine Lung Screening criteria
- Address Social Determinants of Health



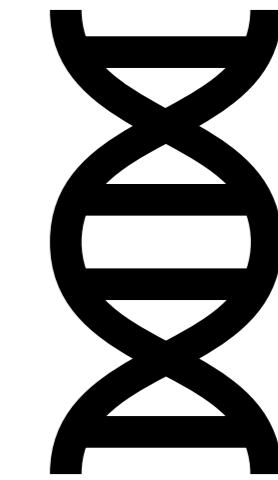
Risk stratification

- Refine NLP referral for review



Nodule Management

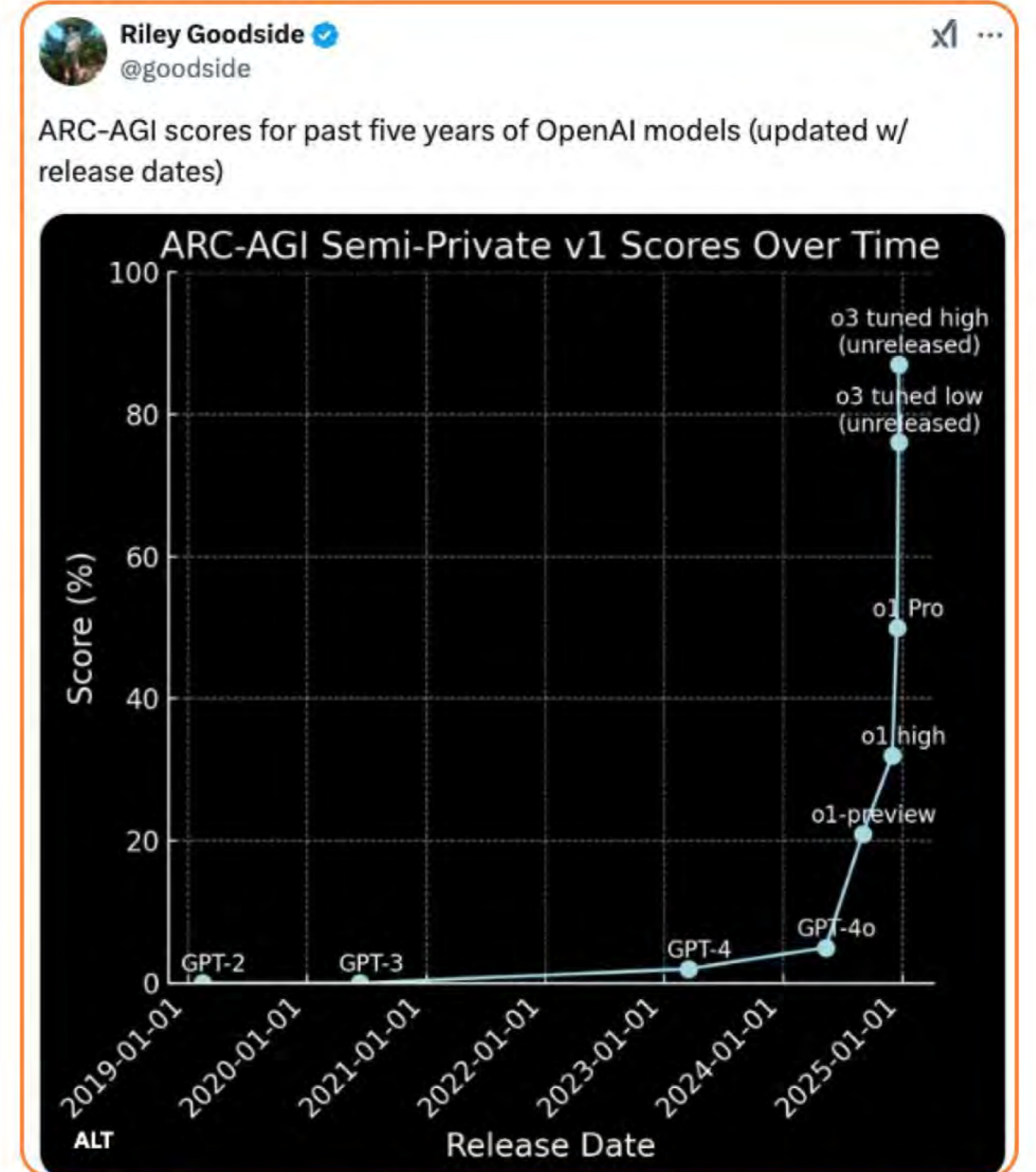
- Improved specificity in nodule detection
- Improved sensitivity for GGOs



Treatment and Outcomes

- Determine best treatment based on all factors known
- **Predict outcomes based on all factors known**

A.I. IS THE WORST IT WILL EVER BE !



FYI, this performance on the ARC-AGI test stirred up some controversy... here's a good recap video on the drama.

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER.

THE FUTURE'S SO BRIGHT, I GOTTA WEAR SHADES



"The Future's So Bright I Gotta Wear Shades",
Timbuk3, 1986, written by Pat MacDonald.
Copyright I.R.S. Records

EARLY DETECTION IS GOOD, CANCER PREDICTION IS BETTER.

APPENDIX

Outcomes From More Than 1 Million People Screened for Lung Cancer With Low-Dose CT Imaging

Gerard A. Silvestri, MD; Lenka Goldman; Nichole T. Tanner, MD, MSCR; Judy Burleson; Michael Gould; Ella A. Kazerooni, MD; Peter J. Mazzone, MD, MPH; M. Patricia Rivera, MD; V. Paul Doria-Rose, DVM, PhD; Lauren S. Rosenthal, MPH; Michael Simanowith; Robert A. Smith, PhD; and Stacey Fedewa, PhD

Conclusion (from abstract):

This study revealed both the positive aspects of CT scan screening for lung cancer and the challenges in implementing screening nationally. Reassuringly, most patients met criteria to be screened. Findings on CT imaging were correlated accurately with lung cancer detection using the Lung-RADS system. A significant stage shift toward early-stage lung cancer was

found. However, **adherence to lung cancer screening was poor, was more likely to occur in the underserved**, and likely contributes to the lower than expected cancer detection rate, all of which will impact the outcomes of patients undergoing screening for lung cancer.



“WE GET ONE CHANCE”

6%

ONLY 6% OF THOSE
ELIGIBLE GET
SCREENED

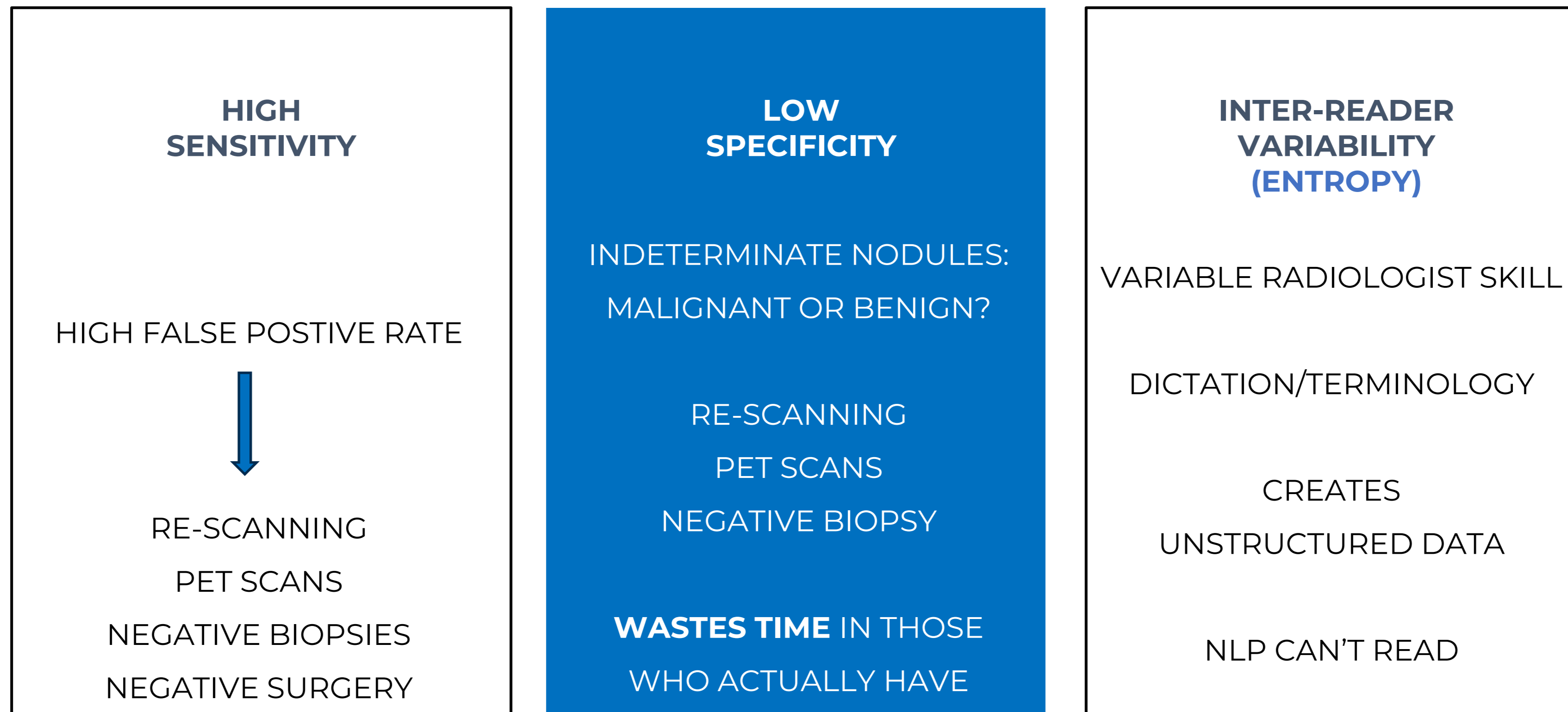
75%

75% WILL **NEVER**
GET A SECOND
SCAN,
WHEN ½ ARE
DISCOVERED

1

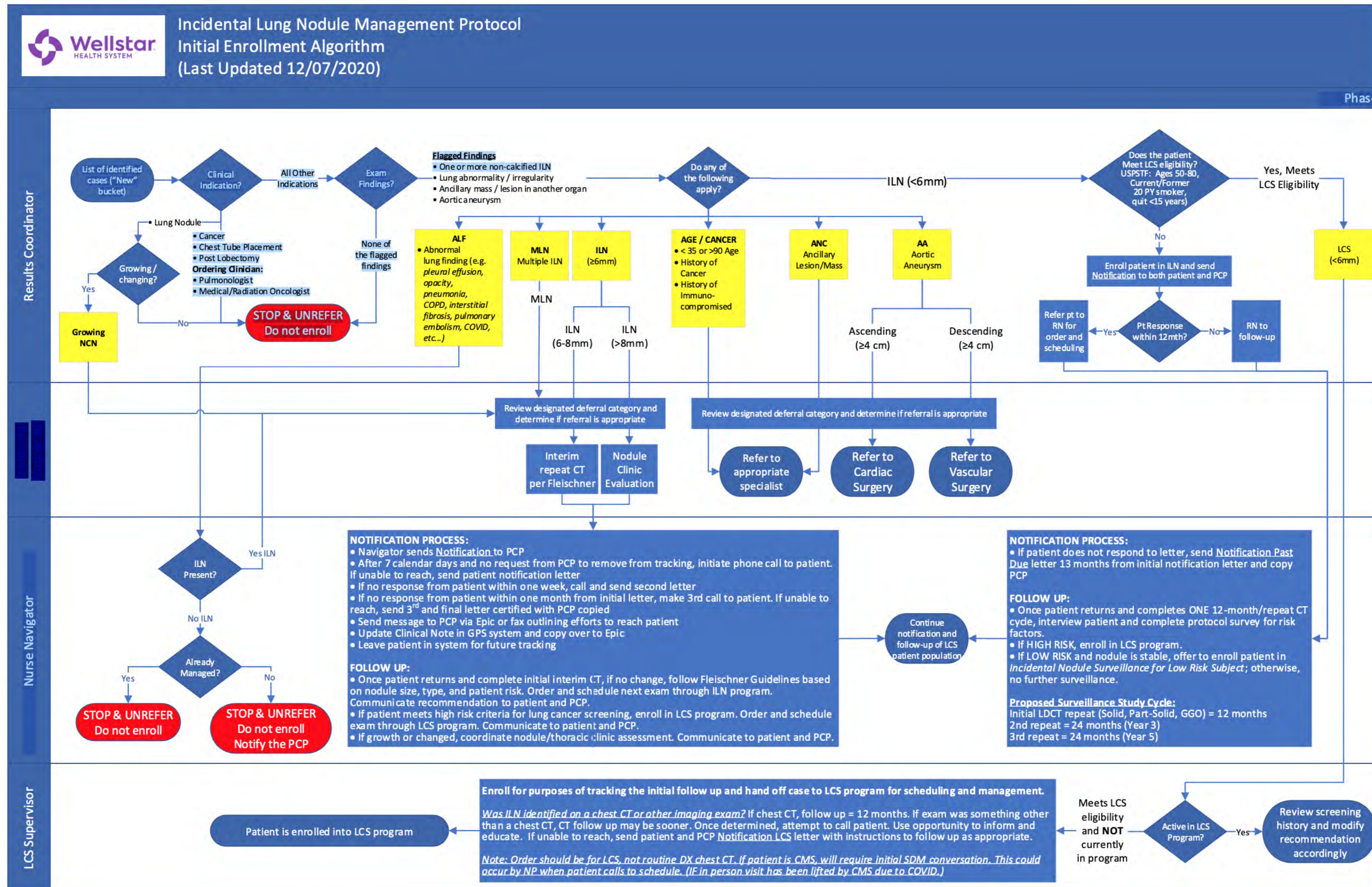
1 SCAN, DONE
FOR ANOTHER
REASON,
DETECTS THE
INCIDENTAL
NODULE

HIGH SENSITIVITY, LOW SPECIFICITY, ENTROPY



INEFFICIENT: TIME AND MONEY AND SURVIVAL WASTED

NO ONE DESERVES TO DIE OF LUNG CANCER



LCS Supervisor

Flowchart Summary:

- Decision: 'Was ILN identified on a chest CT or other imaging exam?'.
 - If 'No', 'Enroll for purposes of tracking the initial follow up and hand off case to LCS program for scheduling and management.' → 'Patient is enrolled into LCS program'.
 - If 'Yes', 'Was ILN identified on a chest CT or other imaging exam?'.
 - If 'chest CT', follow up = 12 months.
 - If 'exam was something other than a chest CT', CT follow up may be sooner. Once determined, attempt to call patient. Use opportunity to inform and educate. If unable to reach, send patient and PCP Notification LCS letter with instructions to follow up as appropriate.
- Decision: 'Meets LCS eligibility and NOT currently in program?'.
 - If 'Yes', 'Active in LCS Program?'.
 - If 'No', 'Continue notification and follow-up of LCS patient population'.
- Decision: 'Active in LCS Program?'.
 - If 'Yes', 'Review screening history and modify recommendation accordingly'.
 - If 'No', 'Continue notification and follow-up of LCS patient population'.

Note: Order should be for LCS, not routine DX chest CT. If patient is CMS, will require initial SDM conversation. This could occur by NP when patient calls to schedule. (If in person visit has been lifted by CMS due to COVID.)