

The Journey to Safe and Effective AI Adoption Gilbert W. Beebe Symposium 2025

Mike Tilkin, CIO and EVP Technology American College of Radiology March 13, 2025

Disclosures

• Employee of the American College of Radiology





Agenda

Background
Recent Provider Perspectives
Navigating the Evolving Al Market
Deploying Al in Production
Notes on LLMs and VLMs











Serve patients and society by empowering members to advance the practice, science, and professions of radiological care



ACR Data Science Institute

Established 2017

Advance data science as core to clinically relevant, safe, and effective radiologic care

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ACR Data Science Institute Established 2017



Christoph Wald, MD, PhD, MBA, FACR ACR Commission on Informatics Chair



Keith J. Dreyer, DO, PhD, FACR ACR DSI Chief Science Officer



Bernardo Bizzo, MD, PhD ACR DSI Associate Chief Science Officer



Woojin Kim, MD ACR DSI Chief Medical Officer

Mike Tilkin

Etta Pisano

Bill Thorwarth



Mike Tilkin ACR Executive Vice President and Chief Information Officer



Laura Coombs, PhD **ACR Vice President of Informatics**



Chris Treml ACR DSI Director of Operations



Bibb Allen Jr., MD, FACR Senior Advisor

DSI Oversight Board

- Christoph Wald
 Laura Coombs
- Bibb Allen
- Keith Drever
- Amy Kotsenas
- Jackie Bello
- Jim Brink
- Howard Fleishon
- Geraldine McGinty

Al Panel Chairs

- Luther B. Adair
- Bhavika Patel
- Robert Rosenberg
- Carlo De Cecco
- Jeremy Durack
- Jay W. Patti
- Alexander M. Norbash
- Christopher T. Whitlow

- Sumit Niogi
- John Tsiouris
- Reid F. Thompson
- Steven Blumer
- Carol Wu
- Alexander J. Towbin
- Adam B. Prater

Senior Scientists

- Sheela Agarwal
- Tarik Alkasab
- Katherine P. Andriole •
- Bernardo C. Bizzo
- Garryi Choy
- Adam E. Flanders
- J. Raymond Geis

- C. Matthew Hawkins
- Jayashree Kalpathy-Cramer
- Matthew P. Lungren
- Gregory N. Nicola
- Chris L. Sistrom
- Curtiland Deville. Jr.
 Neil Tenenholtz

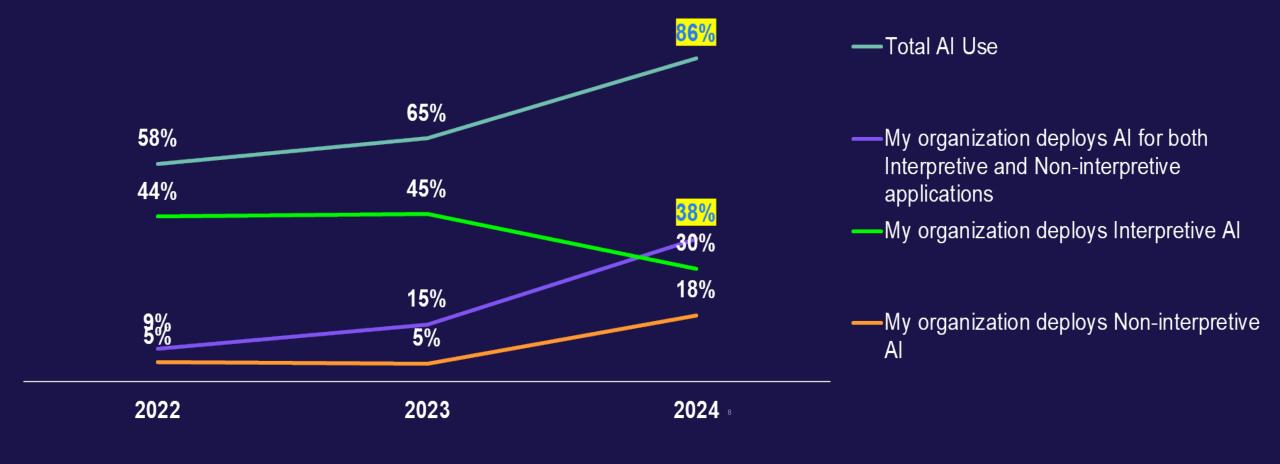


Provider Perspectives on Al Market

Results from a 2024 ACR DSI Al Tracking Survey



Al Adoption in Radiology Continues to Grow



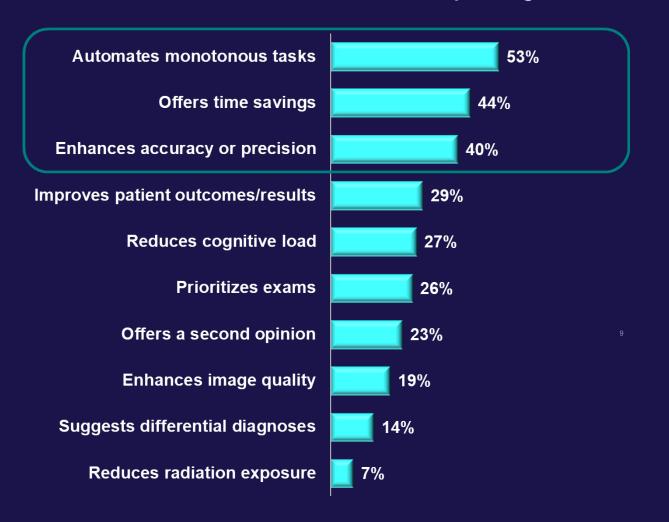
DSI AI Tracking Survey (ACR publication forthcoming)

- Sample of nearly 1,000 (n=922) Members, representative of the population: age/tenure, practice type, geography, gender, specialty
- Data Collected in Fall/Winter 2024/2025 with margin of error (+/-) of 3.2% @ 95% confidence level



Most appealing Attributes of AI: Address volume and efficiency challenges allowing radiologists to focus on higher-impact activities

User – Perceived Benefits of Incorporating Al



The primary value of Al is its role in mitigating the radiology volume crisis – automating repetitive tasks (53%) and saving time (44%) directly address systemic work pressures.

d



Accuracy and Medicolegal liability lead AI Implementation concerns

- Accuracy and liability concerns highlight the need for solutions that:
 - Ensure that Al/ML is accurate and in compliance with safety protocols
 - Provide risk mitigation strategies for those practices adopting Al
 - Highlight those products and services that can be trusted and mitigate medicolegal liability

Concerns about Implementing Al





Most Trust AI outputs moderately or somewhat

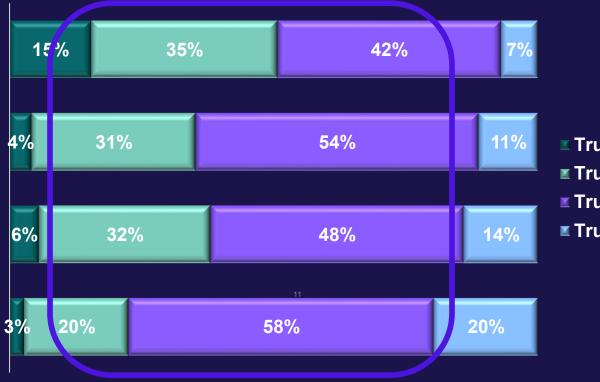
Al output from medical images for reducing acquisition times

Al output from medical images for detection or diagnostic reasons

Al output from medical images for synthetic contrast/non-contrast applications

Al output from medical images for modality transformations

Trust in Al Outputs

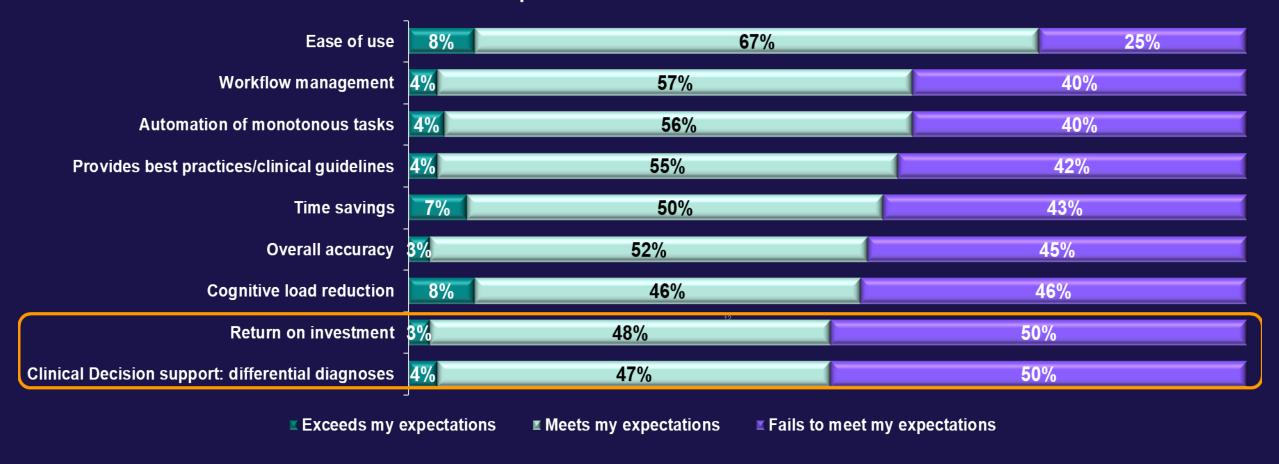


- **In Trust completely**
- **▼ Trust moderately**
- ▼ Trust somewhat
- **I** Trust not at all



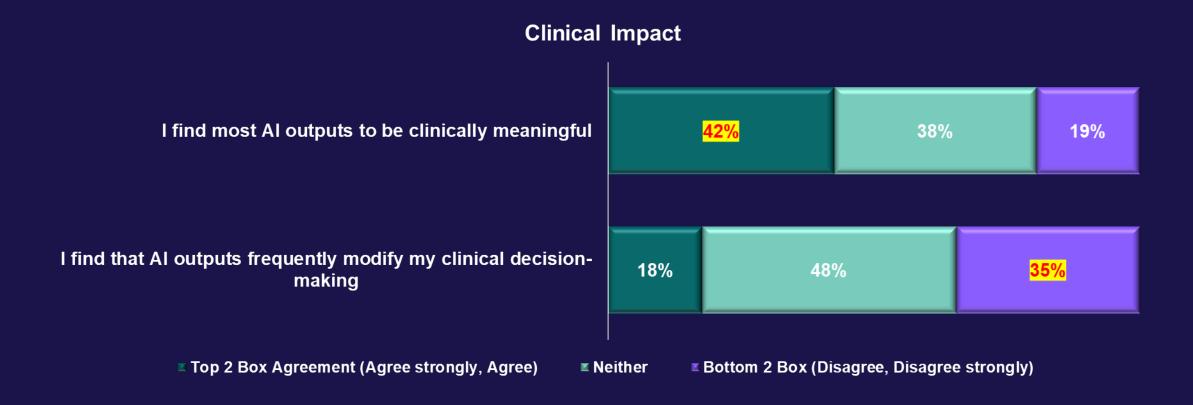
Al Delivers on core expectations; further development needed on ROI and Decision Support to encourage and expand use

Expectations around Al Use





Al Demonstrates clinical value; however, shows limited current influence on clinical decision-making





Adoption and the Al Marketplace

Technology Diffusion and the AI the Market



Identifying the gaps and Research Imperatives

ARTICLE IN PRESS

ORIGINAL ARTICLE

A Road Map for Translational Research on Artificial Intelligence in Medical Imaging: From the 2018 National Institutes of Health/RSNA/ACR/The Academy Workshop

Bibb Allen Jr, MD^a, Steven E. Seltzer, MD^{b,c}, Curtis P. Langlotz, MD, PhD^d, Keith P. Dreyer, DO, PhD^e, Ronald M. Summers, MD, PhD^f, Nicholas Petrick, PhD^g, Danica Marinac-Dabic, MD, PhD, MMSC^b, Marisa Cruz, MD^f, Tarik K. Alkasab, MD, PhD^e, Robert J. Hanisch, PhD^f, Wendy J. Nilsen, PhD^f, Judy Burleson, BSW, MHSA^f, Kevin Lyman, BS^m, Krishna Kandarpa, MD, PhDⁿ

Abstract

Advances in machine learning in medical imaging are occurring at a rapid pace in research laboratories both at academic institutions and in industry. Important artificial intelligence (AI) tools for diagnostic imaging include algorithms for disease detection and classification, image optimization, radiation reduction, and workflow enhancement. Although advances in foundational research are occurring rapidly, translation to routine clinical practice has been slower. In August 2018, the National Institutes of Health assembled multiple relevant stakeholders at a public meeting to discuss the current state of knowledge, infrastructure gaps, and challenges to wider implementation. The conclusions of that meeting are summarized in two publications that identify and prioritize initiatives to accelerate foundational and

- Identifying the gaps and Research Imperatives
- Promoting Standards and Interoperability

ACR DSI Urges AI Standardization, Interoperability and Reportability at NIH Workshop

Share A Recommend Bookmark

American College of Radiology Data Science Institute™ (ACR DSI (2) Chief Medical Officer Bibb Allen, MD, FACR, co-chaired the recent National Institute of Biomedical Imaging and Bioengineering (NIBIB) Workshop on Artificial Intelligence (AI) in Medical Imaging. Proceedings from the workshop will be published as a research roadmap for health care and scientific professionals.

ACR DSI presenters stressed the need for standardized AI use cases to not only ensure that the relevant clinical questions are answered, but to promote interoperability and reportability for ongoing quality improvement — including that:

- . The AI market is dependent upon both the development of AI algorithms and integration with current digital solutions (PACS, reporting systems, etc.)
- · The DSI is working with various clinical experts, academic radiology departments and other radiology professional organizations to create standardized use cases for developing AI algorithms that will solve the most important clinical problems radiologists encounter
- Compared to algorithms developed by industry developers or at single academic institutions, Al solutions that follow DSI Al use case standards will be able to be readily integrated into



ACR DSI Chief Medical Officer Bibb Allen, MD, FACR

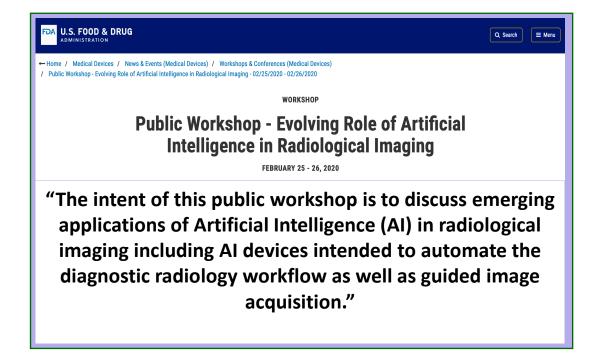


ACR DSI Chief Science Officer Keith Dreyer, DO, PhD, FACR

ACR DSI chief science officer Keith Dreyer, DO, PhD, FACR, specifically outlined the Lung-RADS® Assist: Advanced Radiology Guidance, Reporting and Monitoring use case that was recently chosen as a pilot project by the FDA-funded National Evaluation System for Health Technology Coordinating Center ™ (NESTcc). Dreyer outlined that the Lung-RADS standardization provided a basis for this standardized use case that will:

- · Utilize existing ACR resources with data from multiple institutions to demonstrate the ability to validate algorithm performance prior to FDA clearance
- Facilitate interoperability between reporting and AI developers to generate standardized data in a real-world setting
- Capture real-world data in a national registry to monitor the performance of AI algorithms in clinical practice and enable both facility-level and developer reporting to ensure the algorithm performs as expected in clinical practice

- Identifying the gaps and Research Imperatives
- Promoting Standards and Interoperability
- Working with Regulators on the Limits of Al (e.g., Autonomous Al)



- Identifying the gaps and Research Imperatives
- Promoting Standards and Interoperability
- Working with Regulators on the Limits of Al (e.g., Autonomous Al)
- Educating and Convening

2018 Economics of Artificial Intelligence

2019 Data Access in Healthcare – Implications for the Artificial Intelligence Ecosystem

2020 Monitoring and Evaluating AI: Challenges and Practical Implications

2021 Al in Clinical Practice: Choosing the Right Tools for your Practice and Patients

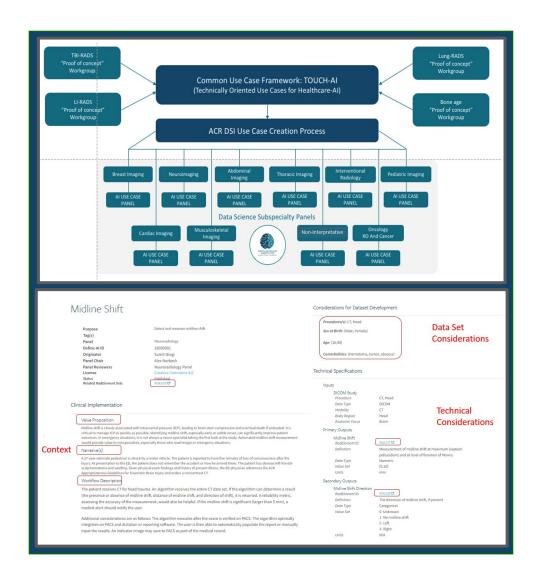
2022 Using AI in Clinical Practice: A Practical Guide for Radiologists

2023 Quantifying the Value of Al: Investment, Costs, Reimbursement and Benefits

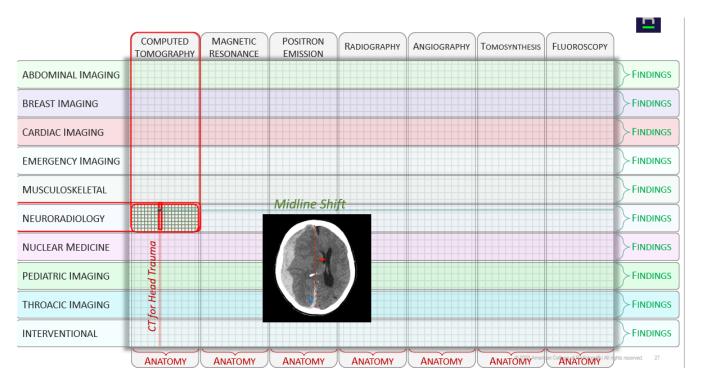
2024 From Data to Diagnosis with Tested, Transparent, and Trustworthy Al

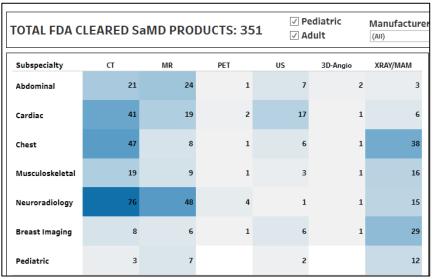


- Identifying the gaps and Research Imperatives
- Promoting Standards and Interoperability
- Working with Regulators on the Limits of Al (e.g., Autonomous Al)
- Educating and Convening
- Defining Use Cases (DEFINE-AI)

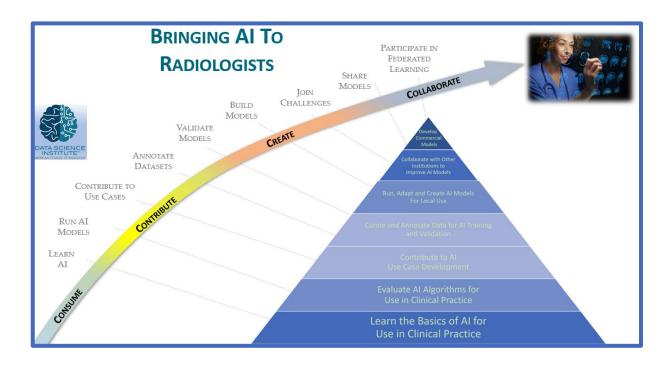


- Identifying the gaps and Research Imperatives
- Promoting Standards and Interoperability
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- Defining Use Cases (DEFINE-AI)





Democratizing AI



- Democratizing Al
- Tools to Foster Engagement & Literacy



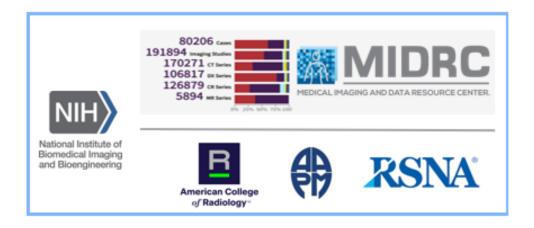
- Democratizing AI
- Tools to Foster Engagement & Literacy
- Research and Data Collection

500+ 100+ 150+ 20,000+ 50+
Clinical Projects Active Clinical Clinical Research Sites Using Conducted Projects Staff TRIAD Research



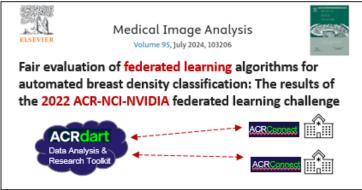


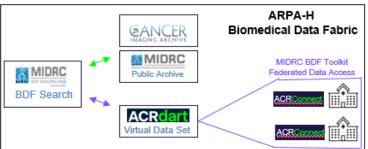




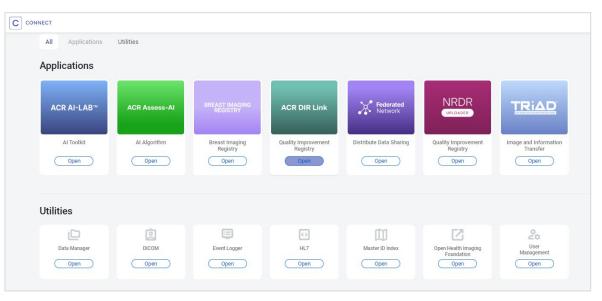
- Democratizing AI
- Tools to Foster Engagement & Literacy
- Research and Data Collection
- Federated Data Access and Learning

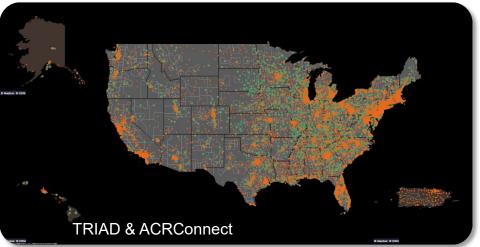






- Democratizing AI
- Tools to Foster Engagement & Literacy
- Research and Data Collection
- Federated Data Access and Learning





- Democratizing AI
- Tools to Foster Engagement & Literacy
- Research and Data Collection
- Federated Data Access and Learning
- Facilitating the Discussion with Regulators, Payers, and Industry



- Democratizing Al
- Tools to Foster Engagement & Literacy
- Research and Data Collection
- Federated Data Access and Learning
- Facilitating the Discussion with Regulators, Payers, and Industry
- Real-world Testing

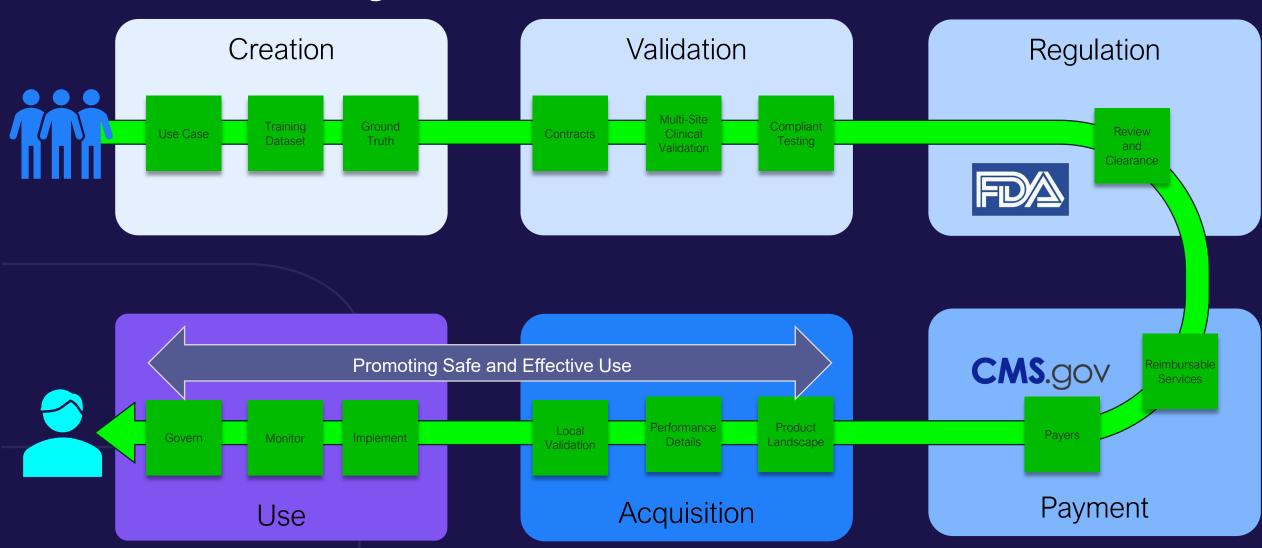


Scaling Adoption

A framework to Promote Safe and Effective Use of Clinical Al



Al Life Cycle



A Central

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Welcom

Products

Manufacturers

Platforms

Current Al Landscape

About

Healthcare Al Challenge

ARCH

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Your Resource for Strategic Al Solutions

Welcome to the ACR Data Science Institute® AI Central database, the most complete and up-to-date online, searchable directory of commercially available Imaging AI products in the United States. Browse through more than 200 FDA-cleared products created by more than 100 manufacturers to find algorithms that best support your patients and workflows.

Search Al Central



Quick Links



Platforms



Transparent Al Products



Assess-Al Compliant



Healthcare Al Challenge

Product Categories









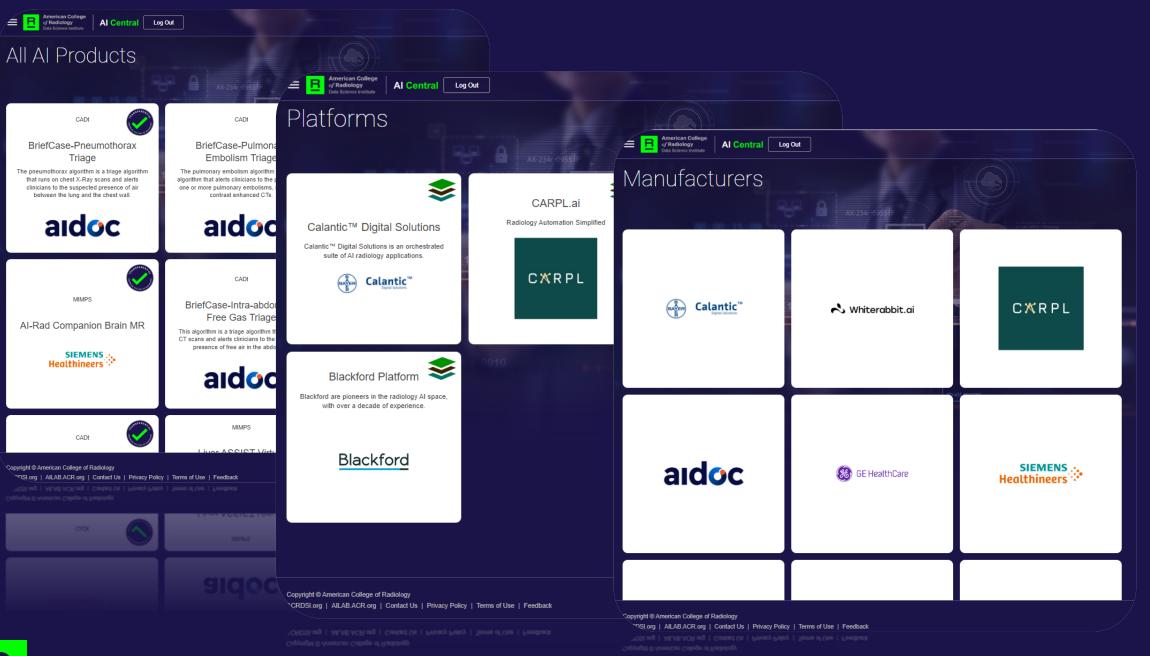




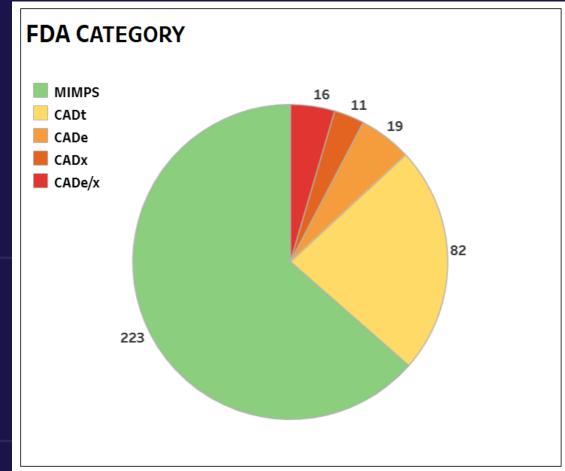


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SaMD Product Drilldown



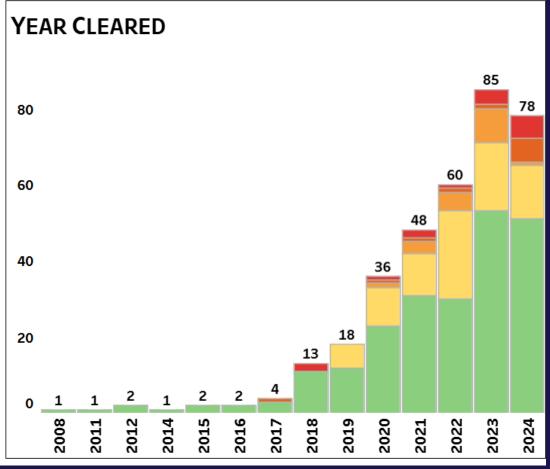
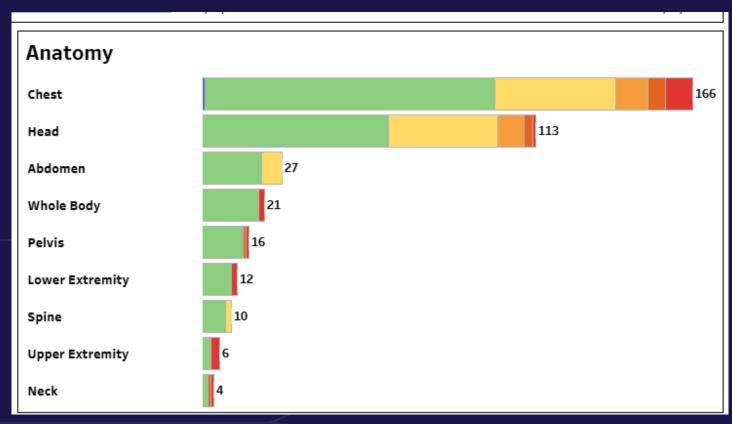


Image Processing And Analysis Tools MIMPS Computer Aided Triage CADt Computer Aided Detection CADe Computer Aided Diagnosis CADx Computer Aided
Detection & Diagnosis
CADe/x

SaMD Product Drilldown



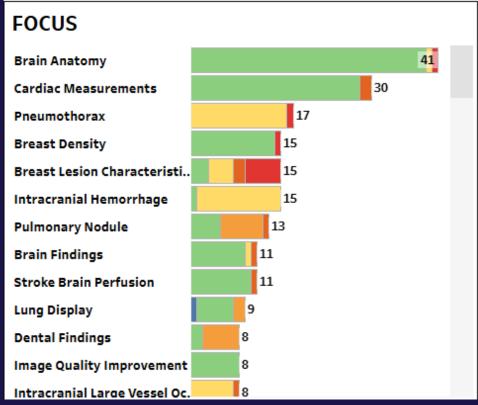


Image Processing And Analysis Tools MIMPS Computer Aided Triage CADt Computer Aided Detection CADe Computer Aided Diagnosis CADx Computer Aided Detection & Diagnosis CADe/x

Promoting Transparency

Transparent-AI Data Elements

Type	Element Name	ID	Description	Required
Model Identifier	Company name	MI.1	Name of company.	Yes
	Product name	MI.2	Name of FDA cleared product.	Yes
	Product version	MI.3	Version of the Algorithm used for inference.	Yes
	Product clearance date	MI.5	Date the product was cleared by FDA.	Yes
	Product FDA submission number	MI.6	Unique FDA ID that is referred as submission number.	Yes
	Additional information	MI.7	Free text field.	No
Model Type	Product category	MT.1	Indicate CAD type the product is cleared for.	Yes
	Additional information	MT.2	Free text field.	No
Model Characteristics	Data inclusion criteria	MC.1	Indicate the inclusion criteria for the product to run and provide output.	Yes
	Data exclusion criteria	MC.2	Indicate the exclusion criteria where the product will not run.	Yes
	Instructions for use	MC.3	Indicate if your company will be able to provide instructions for use document.	Yes
	Additional information	MC.4	Free text field.	No
Indications for Use	Clinical output	IU.1	What is the finding(s) that the product is intended to output	Yes
	Intended user	IU.2	Indicate the intended user of the product.	Yes
	Patient age	IU.3	Age group the product cleared is cleared for.	Yes
	Body area	IU.4	What target body part is the product intended for use.	Yes
	Modality	IU.5	Target modality for the product.	Yes
	Additional information	IU.6	Free text field.	No
Model Performance (performance testing and/or reader study)	Study type	MP.1	Indicate Reader Study or Stand-Alone Performance.	No
	Reference standard (ground truth)	MP.2	What was the reference standard (ground truth) based on?	No

	Number of readers	MP.3	If reader study was conducted, how many readers read each case?	No
	Number of cases	MP.4	How many image studies were used to evaluate model performance?	No
	Number of positive and negative cases	MP.5	Indicate number of positive and negative cases.	No
	Age range	MP.6	Indicate age range for data set.	No
	Sex ratio	MP.7	Indicate sex ratio of data set.	Yes
	Race/ethnicity ratio	MP.8	Indicate the race/ethnicity breakdown of data set.	Yes
	Geographic breakdown	MP.9	Indicate the geographic breakdown of data set.	Yes
	Manufacturer list	MP:10	Indicate the manufacturers in data set.	Yes
	Scanner list	MP.11	Indicate the scanners in data set.	Yes
	Number of sites	MP.12	How many sites were used.	No
	Percent of cases with and without finding	MP:13	Indicate the percent of positive and negative cases.	No
	Model accuracy	MP.14	Indicate the model accuracy.	No
	Model sensitivity	MP.15	Indicate the model sensitivity.	Yes
	Model specificity	MP.16	Indicate the model specificity.	Yes
	Additional information	MP.17	Free text field.	No
Details on Training Data Sets	Model dataset — Number of cases	TD.1	Indicate number of cases the model was trained on.	No
	Age range the model was trained on as well as age distribution	TD.2	Indicate age range the model was trained on as well as the age distribution.	No
	Sex ratio model was trained on	TD.3	Indicate sex ratio the model was trained on.	No
	Positive and negative cases	TD.4	Indicate number of positive and negative cases the model was trained on.	No
	Race/ethnicity ratio	TD.5	Indicate the race/ethnicity break- down the model was trained on.	No
	Geographic breakdown	TD.6	Indicate the geographic break- down the model was trained on.	No
	Manufacturers	TD.7	Indicate the manufacturers the model was trained on.	No
	Scanners	TD.8	Indicate the scanners that the model was trained on.	No
	Additional information	TD.9	Free text field.	No
Model Limita- tions, Warnings and Precautions	Supported scanners/ manufacturer	ML.1	Indicate if algorithm will only work on specific scanner or manufacturer.	No





Assess-Al

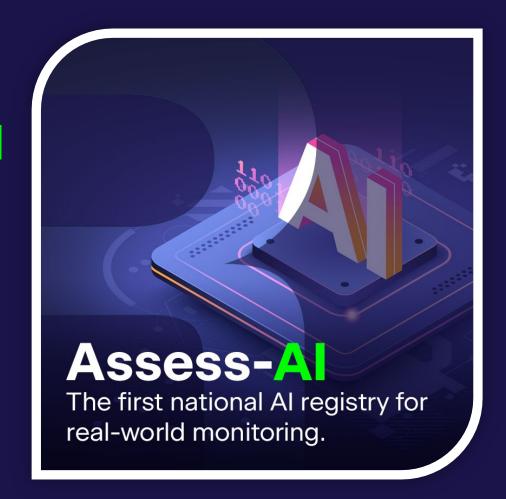
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Assess-Al

Monitoring performance in clinical practice by capturing real clinical data.

First national registry for monitoring Al Results





RSNA ACR RadElement RIS Al Vendor PACS

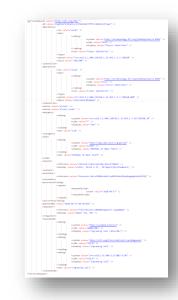
Assess-Al National Registry



Facility



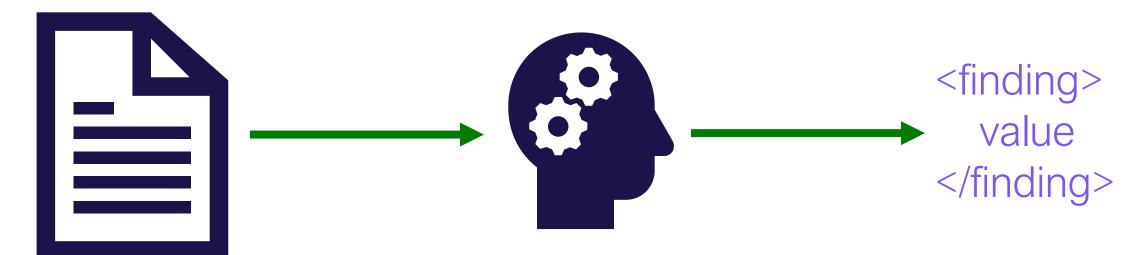






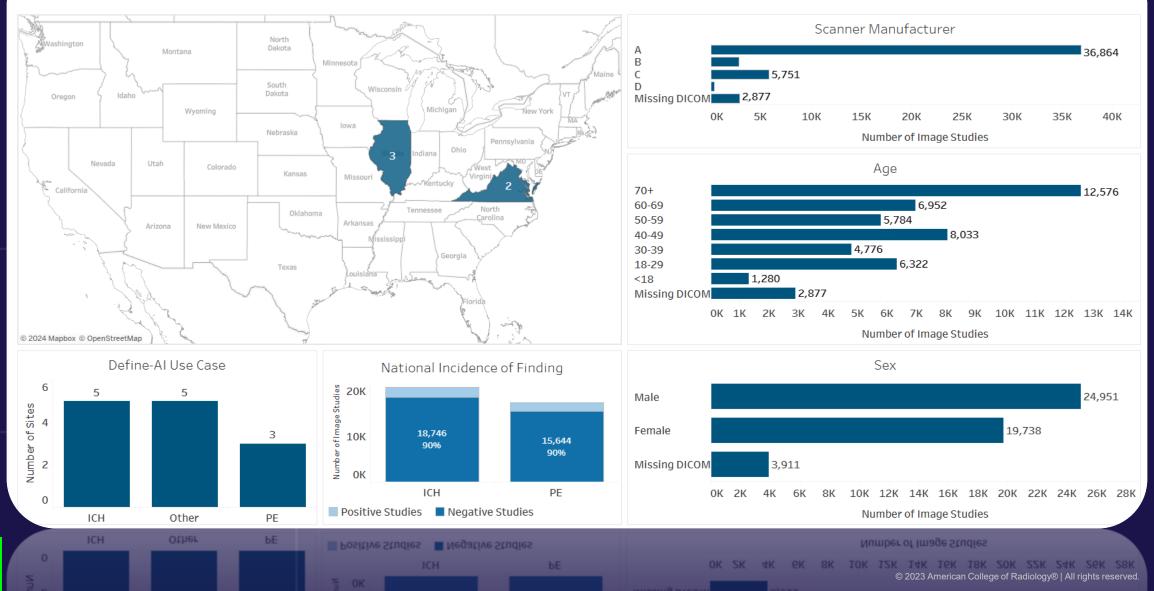


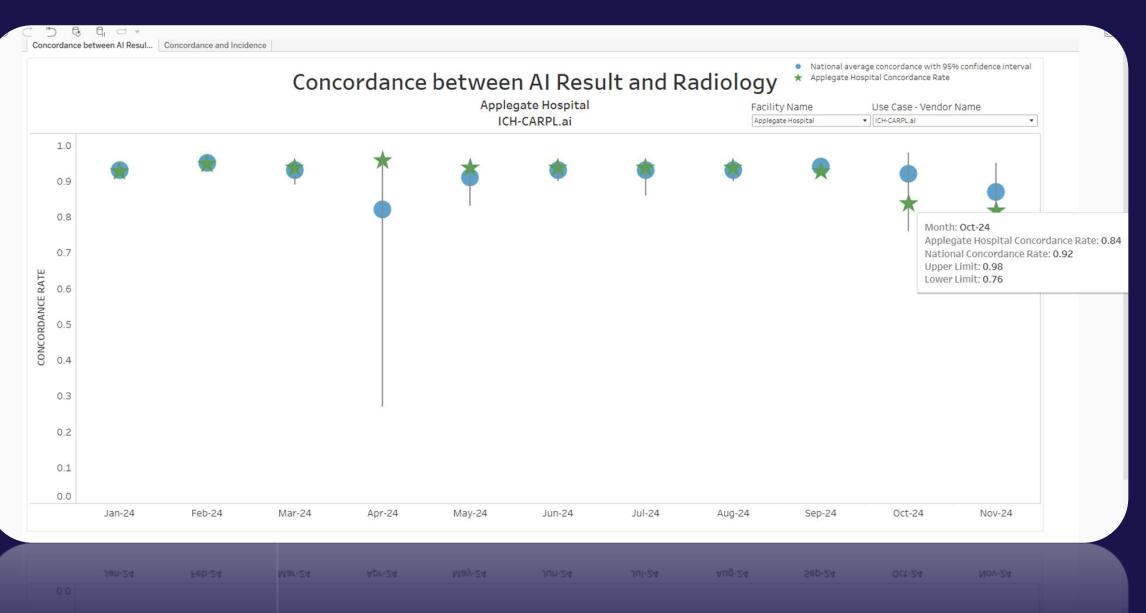


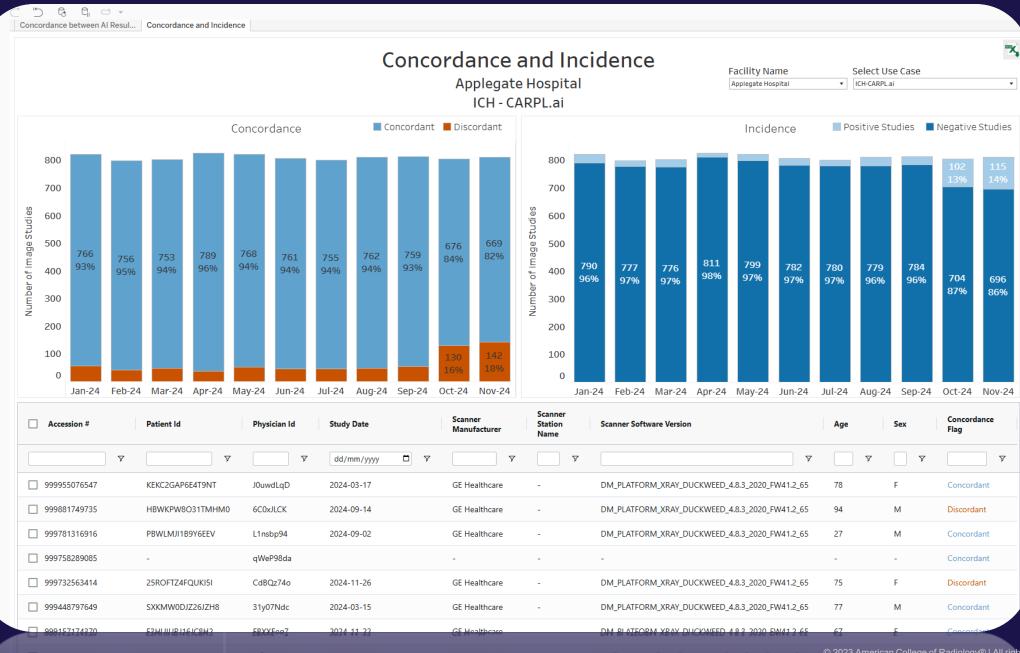




National Assess-AI Registry Dashboard







ARCH-AI

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ARCH-AI

First national artificial intelligence quality assurance program for radiology facilities







Designation







Accreditation















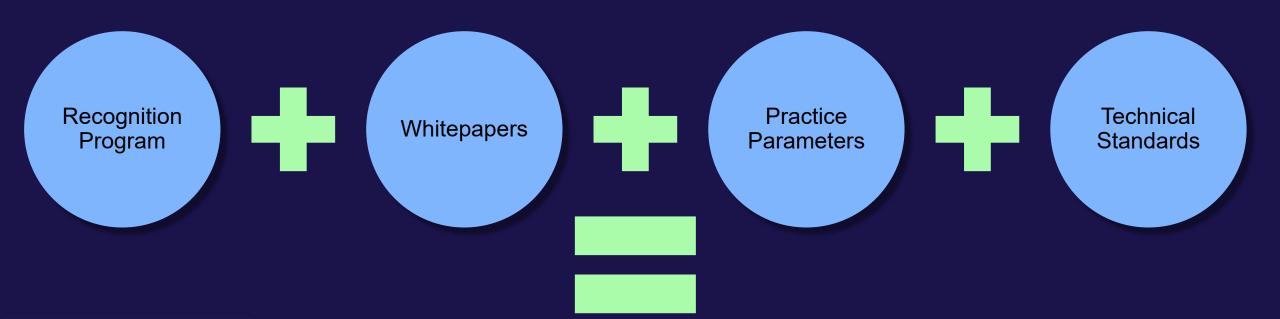


Recognition













Articles

Publish

Topics

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ORIGINAL ARTICLE · Articles in Press, March 07, 2025

The Road Map for ACR Practice Accreditation for Radiology Artificial Intelligence

David B. Larson, MD, MBA $\stackrel{\triangle}{\sim}$ $\stackrel{\square}{\boxtimes}$ · Mythreyi Bhargavan-Chatfield, PhD b · Michael Tilkin, MS c · Laura Coombs, PhD ^d · Christoph Wald, MD, PhD, MBA ^e

Affiliations & Notes ✓ Article Info ✓

















ow Outline

Abstract

As the use of artificial intelligence (AI) continues to grow in radiology, it has become clear that its real-world performance often differs from that demonstrated in premarket testing, underscoring

Article n

Metric dat

Criteria

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Governance Selection Testing Monitoring

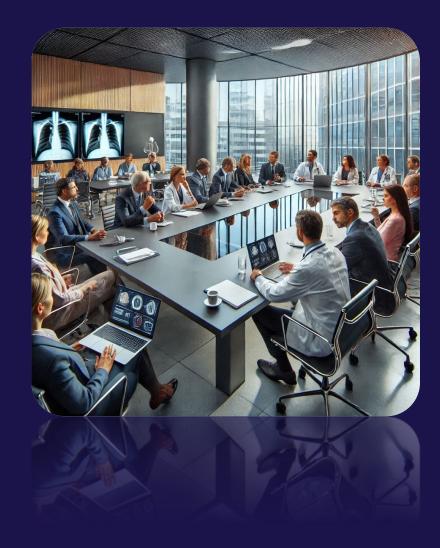


Governance

- Establish an Al governance group to develop policies for evaluating, accepting, monitoring, and sunsetting models
 - Clinical champion (radiologist)
 - Technical lead
 - Administrative personnel
 - Data scientists
 - Informaticists
- Maintain a local inventory of clinically deployed model(s).
 - Model Version
 - Deployment Date
 - Instructions For Use

 Mandatory checkpoint with local Cybersecurity and/or Compliance teams before the deployment





Model Selection

 Consulting with Al Central (or similar) for selecting Al models

Document the inclusion criteria or DICOM routing rules

Document the purpose of the Al model



```
def is_head_ct(dicom_file):
    """
    Check if the DICOM file is a head CT scan.
    """
    ds = pydicom.dcmread(dicom_file)
    if ds.Modality == "CT" and "HEAD" in ds.SeriesDescri
        return True
    return False
```

Midline Shift

Purpose Detect and measure midline shift
Tag(s)
Rapol Neuroradiology

Panel Neuroradiology
Define-AI ID 18030001
Originator Sumit Niogi
Panel Chair Alex Norbash
Panel Reviewers Neuroradiology Panel
License Creative Commons 4.0

Status Related RadElement Sets

Clinical Implementation

Value Proposition

Midline shift is closely associated with intracranial pressure (ICP), leading to brain stem compression and eventual death if untreated. It is critical to manage ICP as quickly as possible. Identifying midline shift, especially early or subtle onset, can significantly improve patient outcomes. In emergency situations, it is not always a neuro-specialist taking the first look at the study. Automated midline shift measurement would provide value to nonspecialists, especially those who read images in emergency situations.

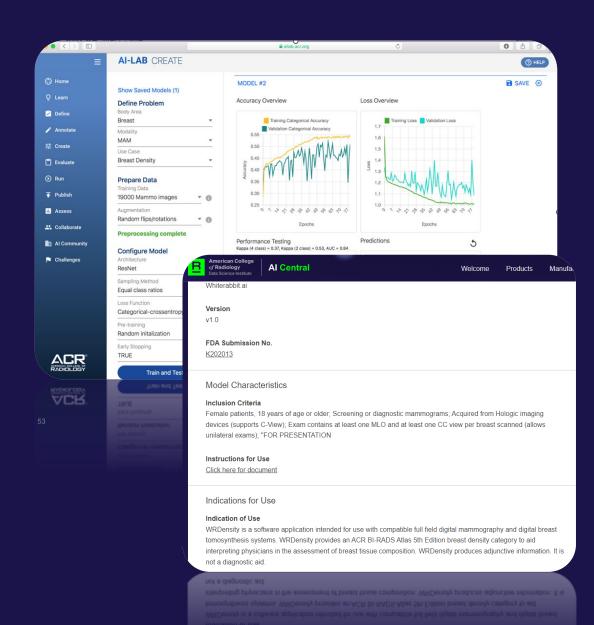


Acceptance Testing

 Document process to assess model performance on local data

Document training procedures for users of the model

 Disseminate information (e.g., Instructions for Use) to local users





Typical Deployment Pathways

Research or non-clinical

- Non-clinical Setting
- Typically, Retrospective Data

Shadow Mode

- Clinical Setting
- Retrospective or Prospective Data
- Test in Production Conditions

Limited (aka Canary)

- Limited Use
- Prospective Data
- Well-Defined Validation Plan
- Usability Feedback

Full Rollout

- Full Production
- Prospective Data
- Production Monitoring

Resources

"Addressing the Challenges of Implementing Artificial Intelligence Tools in Clinical Practice: Principles From Experience"

Bizzo, et al., Journal of the American College of Radiology Volume 20, March 2023

"Implementation of Clinical Artificial Intelligence in Radiology: Who Decides and How?" Dania Daye, MD, et al, PhD, Radiology: Volume 305: Number 3—December 2022

Monitoring

Local monitoring

Participation in registry







Locally Developed Al

Trusted personnel

Follows standard procedures







Beth Israel Lahey Health Lahey Hospital & Medical Center





Yale New Haven Hospital







Department of Radiology and Imaging Sciences

















Atlantic Health System











UNIVERSITY OF WASHINGTON MEDICAL CENTER



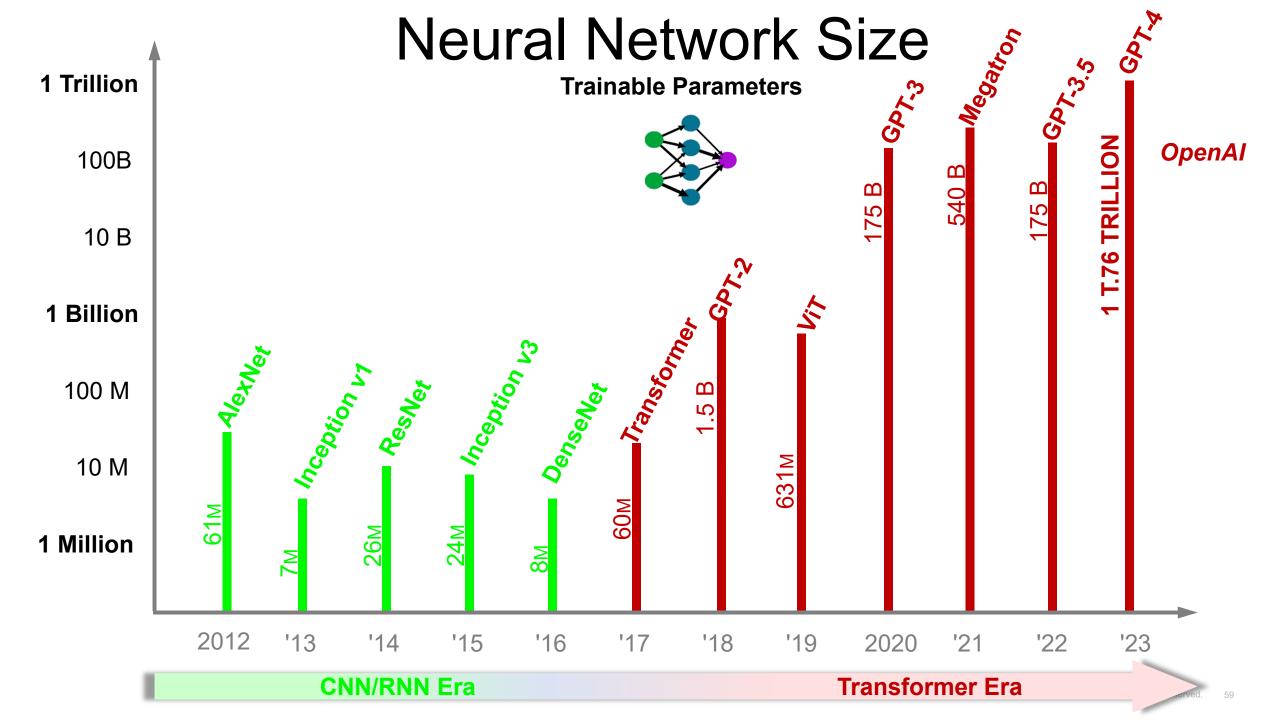




Notes on LLMs and GenAl

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Simple vs. Complex Tasks







General	VIEW DOCS
PREDICTED CONCEPT	PROBABILITY
dog	0.999
mammal	0.999
canine	0.997
pet	0.994
animal	0.987
one	0.985
portrait	0.983
no person	0.967
cute	0.957
domestic	0.950







VIEW DOCS
PROBABILITY
0.987
0.984
0.981
0.962
0.952
0.945
0.937
0.926
0.896
0.886
0_870

Focused.

rve







ChatGPT

In the picture, there is a dog standing in front of a green chair. The chair's cushion appears to have been torn apart, with the stuffing scattered all around on the floor, suggesting that the dog may have destroyed it. The dog has a guilty or sheepish expression, which is commonly interpreted as a sign of knowing it did something wrong. The pillow from the chair is on the floor, undamaged, which might imply that it was removed before or after the cushion was torn apart.

what is happening in this picture



In the picture, there is a dog standing in front of a green chair. The chair's cushion appears to have been torn apart, with the stuffing scattered all around on the floor, suggesting that the dog may have destroyed it. The dog has a guilty or sheepish expression, which is commonly interpreted as a sign of knowing it did something wrong. The pillow from the chair is on the floor, undamaged, which might imply that it was removed before or after the cushion was torn apart.







Define

Annotate

₹ Create

Evaluate

Run

→ Publish

II. Assess

24 Collaborate

Al Community

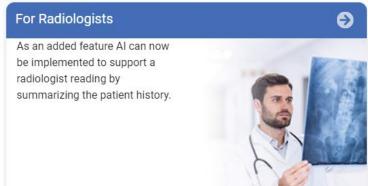
Generative AI

Challenges

Generative AI: Radiology Meets ChatGPT

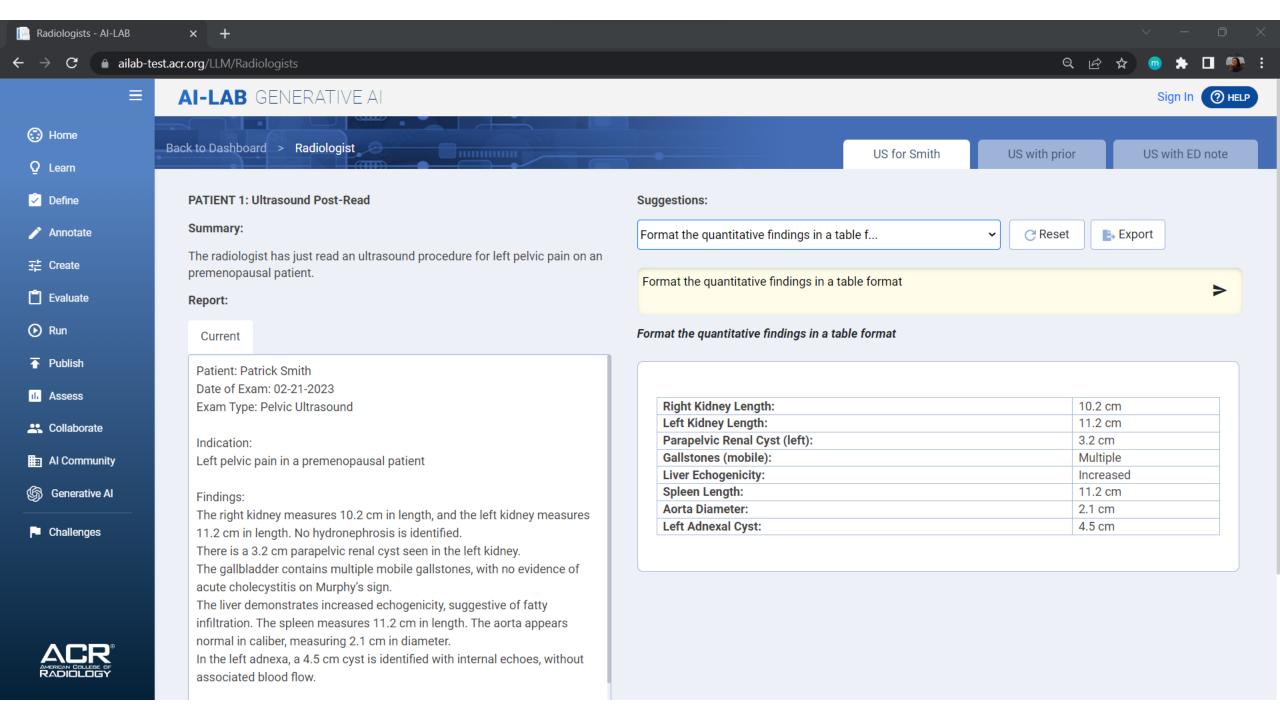
Welcome to ACR's GPT reference, research and translation tool. Our GPT assisted tool will help patients, physicians and radiologists discover previously hard to find documentation, definitions and reports concerning a multitude of imaging procedures.

















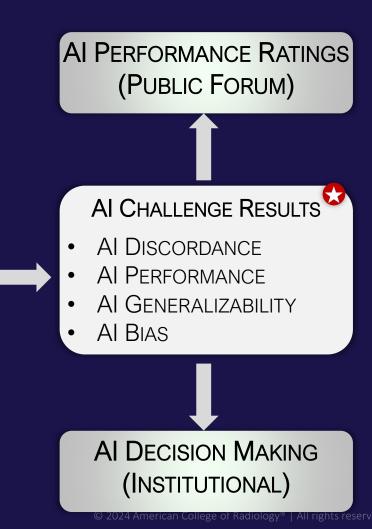


UW Health UW Medicine



A Collaborative Community for Healthcare Al Validation and Monitoring

- Geographically & Demographically Diverse, Embargoed Data
- 2 Unlimited Clinical Use Cases (e.g. Draft Report Generation)
- Geographically & Institutionally Diverse Experts
- 4 Expert evaluation of Al Agents with Input Data





Powered by MGB Al Arena

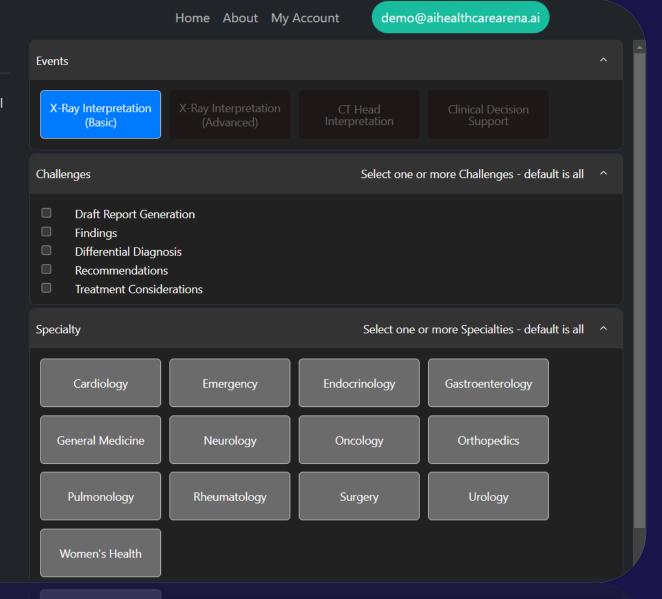
Description

Analytics

The X-Ray Interpretation Event includes radiography cases from major body parts, organized by Clinical Specialties.

Healthcare professionals can review and rate responses from foundation models that process radiography cases for various Challenges.

Each response should be rated on a five-point clinical skill level scale: Radiologist, Fellow, Resident, Medical Student, or Unacceptable.



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Draft Report Generation

Challenge 1 of 1

Findings:The lumbar spine X-rays in the anteroposterior and lateral views show normal alignment of the vertebral bodies. There is no evidence of fracture or dislocation. The intervertebral disc spaces are preserved. There is no significant osteophyte formation or other signs of degenerative changes. The sacroiliac joints appear normal. No abnormal soft tissue calcifications are noted.

Impression:Normal lumbar spine X-ray. No acute osseous abnormalities or significant degenerative changes are observed.

Recommendations:No immediate follow-up is necessary based on the current imaging. Clinical correlation is recommended to determine if further evaluation is needed based on symptoms. If symptoms persist or worsen, consider advanced imaging such as MRI for further assessment.

Select the clinical skill level you would assign to the model's response

Attending (Outstanding)

Fellow (Good)

Resident (Acceptable)

Student (Poor)

Unacceptable

Next Case

Next Cas

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Images

All Models



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The above answer was generated by OpenAI (GPT 40 mini)

Enter Comments (Optional)

Save

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Fellow (Good)

Resident (Acceptable)

Student (Poor)

Unacceptable

Next Challenge

New Case

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New Case

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Images

All Models

Anthropic (Haiku) Anthropic (Opus)

Anthropic (Sonnet)

Gemini (Flash)

Gemini (Pro)

OpenAl (GPT 4o)

OpenAl (GPT 4o mini)

Harrison (RAD1)

Microsoft (Draft CXR)

Instruction Prompt Unformatted Result Formatting Prompt Result

Findings:The images show multiple vertebral bodies with irregularity and decreased attenuation, suggesting spinal degenerative changes. There is also evidence of mild scoliosis with curvature of the spine. The spinal canal appears patent without significant compromise. The soft tissues appear unremarkable.Impression:The findings are consistent with age-related degenerative changes of the spine, including spondylosis and mild scoliosis. There is no evidence of acute injury or significant spinal canal narrowing.Recommendations:No immediate intervention is required. The patient should continue with conservative management, including physical therapy and pain management as needed. If symptoms persist or worsen, further imaging evaluation, such as MRI, may be considered to assess for any underlying pathology.

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Draft Report Generation

Challenge 1 of 1

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don't count my vote

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Instruction Prompt Unformatted Result Formatting Prompt Result

* You are a Radiologist, create a draft report for the images that contains the findings, a draft impression, and any follow-up recommendations. Put them all in separate paragraphs and exclude any patient information. Do not include any disclaimers as this is for educational purposes only.

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Challenge 1 of 1

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Next Challenge

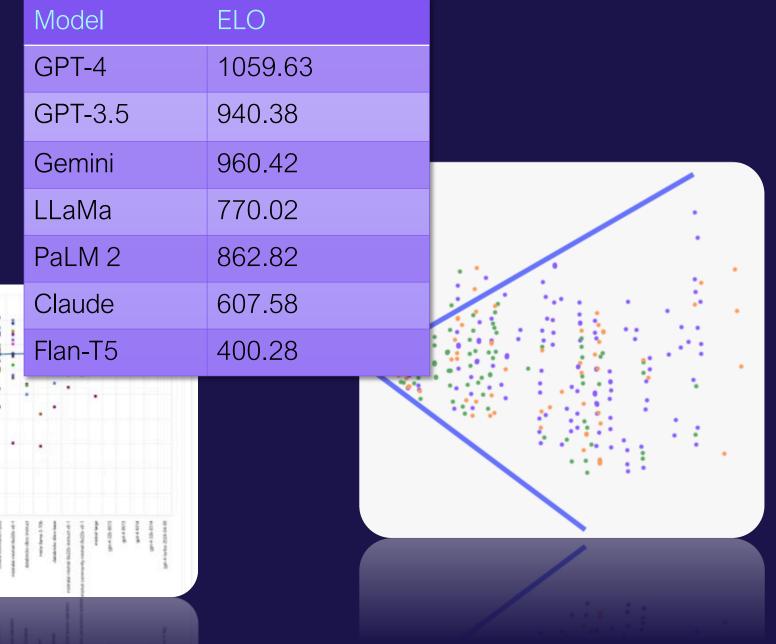
New Case

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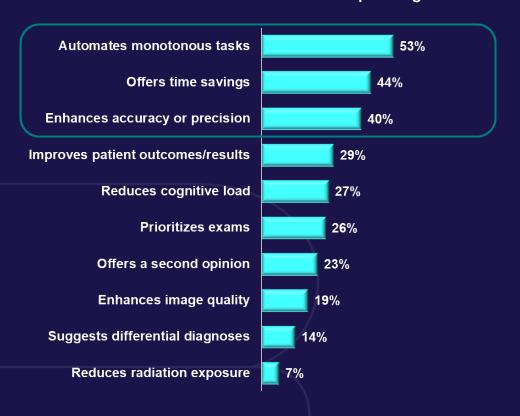


Conclusion



Risk vs. Reward: Factors Influencing the Calculation

User - Perceived Benefits of Incorporating Al





Factors	Examples of Influences
Subjective Risk Attitudes	Individual, Practice, Local Conditions/Market
Type of Intended Use	Triage, Detection, Diagnosis
Clinical Use Case	Different clinical use cases influence risk attitudes
Operational Safeguards	Transparency, Monitoring, Best Practice Implementation



Thank You



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