



Mahidol University

Faculty of Medicine Ramathibodi Hospital

Department of Clinical Epidemiology and Biostatistics

Key Considerations Before AI Deployment in Healthcare

Tint Lwin Win

Student | Data Science in Healthcare and Clinical Informatics

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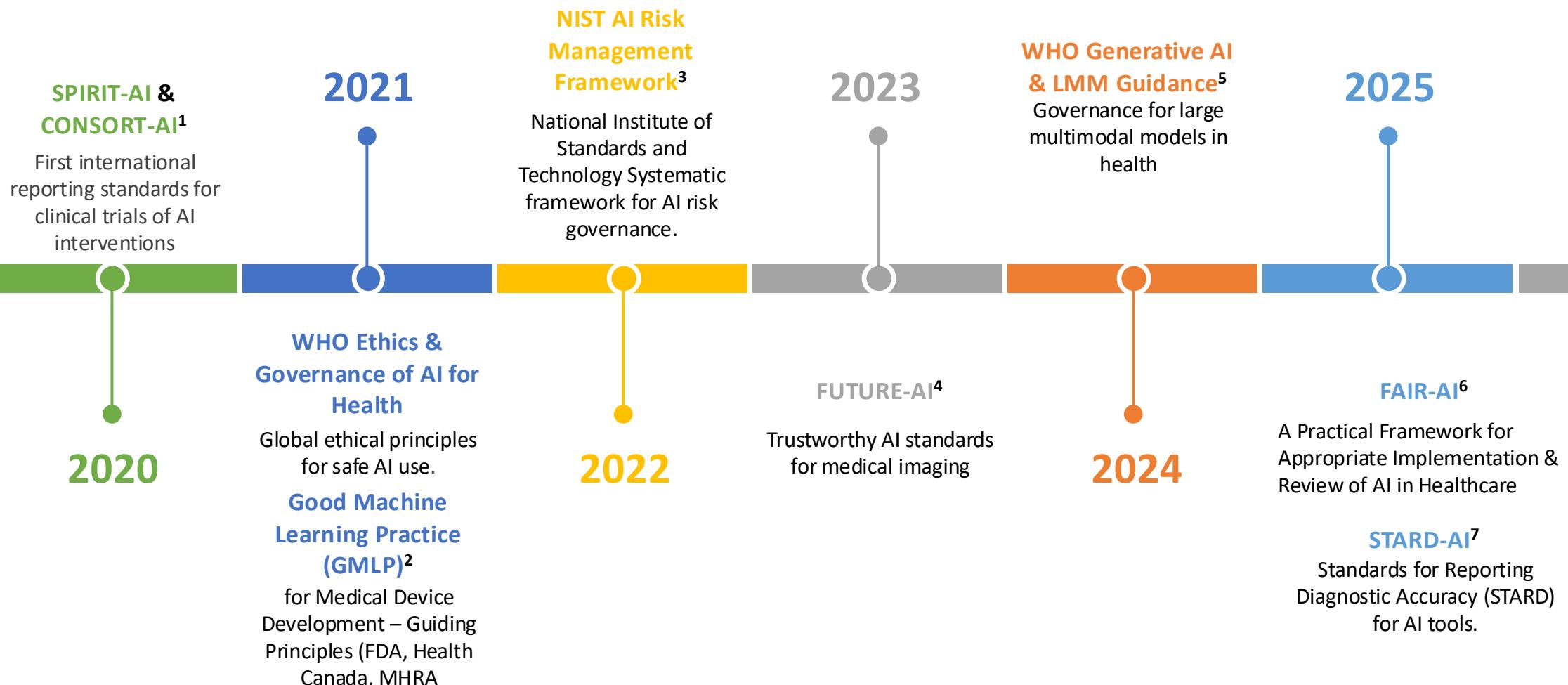
Balancing Innovation and Risk in Clinical AI

- Rising adoption of AI in clinical practice.
 - Potential Benefits:
 - Improved outcomes, efficiency, reduce cost.
 - Potential Risks:
 - Health inequities, patient harm.
- Need for **governance and risk controls**.
 - To ensure AI is **beneficial, safe, and equitable**.





AI-in-Healthcare Frameworks



[1] Ibrahim, H., Liu, X., Rivera, S. C., Moher, D., Chan, A. W., Sydes, M. R., Calvert, M. J., & Denniston, A. K. (2021). Reporting guidelines for clinical trials of artificial intelligence interventions: the SPIRIT-AI and CONSORT-AI guidelines. In *Trials* (Vol. 22, Issue 1). BioMed Central Ltd. <https://doi.org/10.1186/s13063-020-0495-1>

[2] Diamond, & Matthew. (2021). *Good Machine Learning Practice for Medical Device Development: Guiding Principles*.

[3] Tabassi, E. (2023). *Artificial Intelligence Risk Management Framework (AI RMF 1.0)*. <https://doi.org/10.6028/NIST-AI.100-1>

[4] Le kadir, K., Frangi, et. al. (2025). FUTURE-AI: International consensus guidelines for trustworthy and deployable artificial intelligence in healthcare. *BMJ*. <https://doi.org/10.1136/bmj-2024-081554>

[5] Ethics and governance of artificial intelligence for health. Guidance on large multi-modal models. Geneva: World Health Organization; 2024. Licence: CC BY-NC-SA 3.0IGO.

[6] Wells, B. J., et. al. (2025). A practical framework for appropriate implementation and review of artificial intelligence (FAIR-AI) in healthcare. *Npj Digital Medicine*, 8(1). <https://doi.org/10.1038/s41746-025-01900-y>

[7] Sounderajah V, Ashrafin H, Golub RM, Shetty S, De Fauw J, Hooft L, Moons K, Collins G, Moher D, Bossuyt PM, Darzi A. Developing a reporting guideline for artificial intelligence-centred diagnostic test accuracy studies: the STARD-AI protocol. *BMJ open*. 2021 Jun 1;11(6):e047709.



The Problem: Missing practical guidance

- Current frameworks are often:
 - Non-binding or too theoretical
 - Lacking practical guidance for real-world implementation
 - Inflexible regarding 'high-risk' definitions in healthcare contexts
- **Goal:** To create a standardized, repeatable, and transparent framework.



A framework for appropriate implementation and review of artificial (FAIR-AI)¹

- It was proposed in 2025
- System-level framework for safe, ethical, and appropriate AI deployment.
- Focuses on readiness, risk assessment, and structured review.
- Provides practical guidance for real-world implementation.



Methodology: How FAIR-AI was built

Review

Narrative
Review:
Synthesized **best
practices**
(Validity, Equity,
Transparency).

Interviews

Stakeholder
Interviews: Leaders,
developers,
healthcare
providers, and
patients.

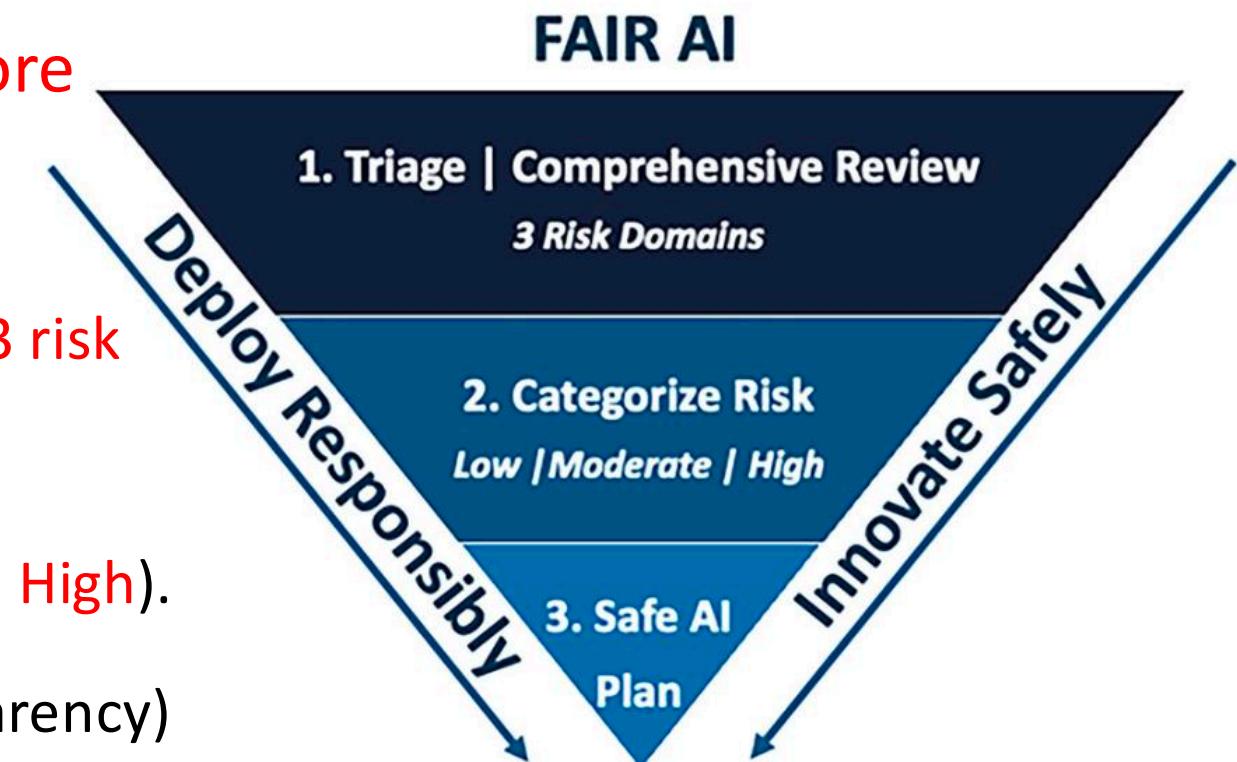
Workshop

Design Workshop:
Consensus from **33**
**multidisciplinary
experts** (individuals with
expertise in legal affairs, regulatory
compliance, cyber security, ethics,
clinical care, clinical informatics,
data science, and research)



Overview: the triangle of responsibility

- The framework is built on **three core components**
 1. Triage & Comprehensive Review (**3 risk domains**).
 2. Categorize Risk (**Low | Moderate | High**).
 3. Safe AI Plan (monitoring & transparency)





Foundational Elements

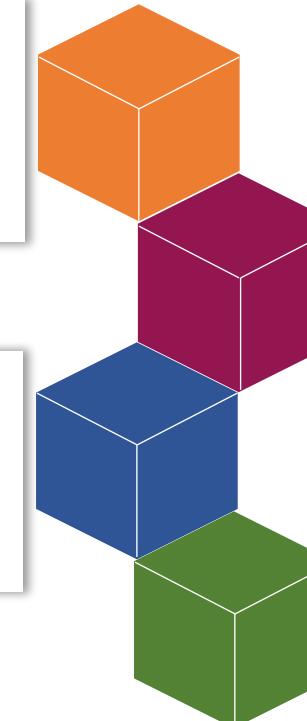
Required before implementation of AI tools in healthcare

Artifacts

Ethics statement & guiding principles.

Escalation Body

Multidisciplinary Governance Committee



Personnel

Accountable data science team

Inventory Tool

'Single source of truth' catalog.



Scope: What Goes Through FAIR-AI?

- **Inclusion**

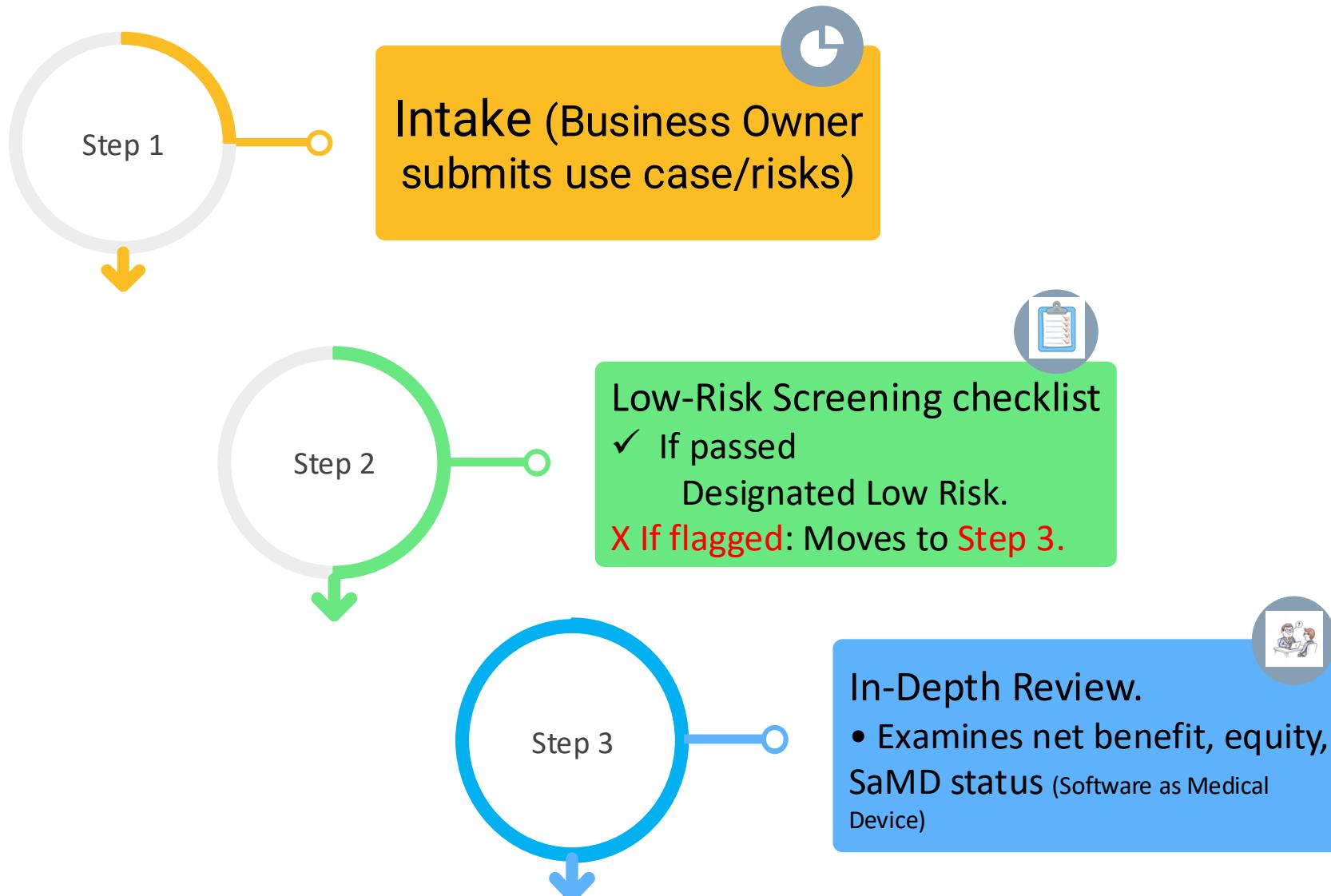
- Broad definition of AI: any computer system capable of activities normally associated with human cognitive effort

- **Exclusions**

- Simple scoring systems & rules-based tools
- Physical medical devices (e.g. FDA regulated devices)
- IRB-approved research protocols



The FAIR-AI review process flow





Risk categories as determined by FAIR-AI evaluation and escalation to AI Governance

Risk Category	Data Science Team	AI Governance Committee
	Evaluation	Escalation Determination
Low	All low risk criteria passed	n/a
Medium	No high risk criteria on in-depth review	n/a
High	Any high risk criteria, and/or supporting evidence is lacking	Proceed under high risk conditions Proceed as high risk - pilot or research required Do not proceed - intolerable risk



Transparency: The AI Label

High-risk solutions require an 'AI Label' for end-users.

AI Label
AI Solution Name
Inventory ID
Developer
Business owner
Model type
FAIR AI Summary
Risk category & initial review date
Low, Medium, High, Uncertain
<ul style="list-style-type: none">• Last review date• Intended use• Directions for interpretation of AI solution output• Performance• Known limitations• Potential risks• FDA approval, if applicable (include type)



Pre-Implementation: The In-Depth Review

1. Technical Evaluation

- Validation evidence & FDA status.

2. Ethical & Equity Analysis

- Bias screening & data representativeness.
- Explainability of the "Black Box".

3. Workflow & Safety

- **Net Benefit:** Confirming utility exceeds risk.
- **Human Oversight:** Ensuring a "human-in-the-loop" for critical decisions.



Safe AI Plan: Post-Deployment Continuous Monitoring

- AI models drift and workflows change over time.
- **Requirement:** Periodic attestation by business owner.
 - **Confirm alignment with intended use case.**
 - **Confirm data/workflows haven't changed.**
 - **Confirm expected benefits are realized.**



Limitations of FAIR-AI

- **Not a regulatory** framework.
- **Resource Intensity:** Requires significant time, personnel, and leadership support, training required for evaluators and end-users.
- **Evaluator Dependency:** Framework relies on the diligence and expertise of reviewers
- **Effectiveness depends** on local implementation.



Real-world AI deployment (The ASSURE Study)

- **The Workflow:** A "Multistage" approach utilizing **Human-in-the-loop**.
 - **Step 1:** Radiologist reads with AI support.
 - **Step 2 (SafeGuard):** If the Radiologist says "Normal" but AI says "High Risk," the case is routed to a **second human expert**.
- **Scale:** Comparison of 208,891 AI-assisted exams vs. 370,692 Standard of Care exams across 109 US sites.
- **Key Results:**
 - **Cancer Detection Rate (CDR):** Increased by **21.6%**
 - **Precision (PPV1):** Increased by **15.0%** (More cancers found without excessive false alarms)
 - **Equity:** Improvements were consistent across Black, Hispanic subpopulations



Evaluating the ASSURE Study through the FAIR-AI Framework

- Based on the **FAIR-AI framework**, the **DeepHealth Breast AI deployment** in this healthcare scenario would be categorized as **Moderate Risk**, because
 - This AI tool is not autonomous and is used as a **clinical decision support** system as it requires **radiologist approval** and is subject to **human oversight**.
 - The **radiologist's oversight** and the **SafeGuard Review** process mitigate the risk, but **false negatives** potentially delay the cancer diagnosis.



Evaluating the ASSURE Study through the FAIR-AI Framework

- Good Practices
 - **Safety Net:** The "SafeGuard Review" directly addresses FAIR-AI's "**Human Oversight**" requirement. The AI cannot unilaterally reject a cancer diagnosis; it triggers a human safety check.
 - **Equity Validation:** They explicitly validated against vulnerable subgroups (race/age) to ensure the model didn't widen disparities, satisfying the **FAIR-AI Equity screen**.
 - **Net Benefit:** Demonstrated clear clinical utility (**increases in CDR/PPV**) **outweighing** the modest **increase in recall rate**.



Evaluating the ASSURE Study through the FAIR-AI Framework

- Areas for Improvement

- The study operated under a "**waiver of consent**". It is unclear if patients were notified that **AI was assessing their cancer risk**.
- The paper lacks detailed reporting on **governance structure** and oversight committees.
- No explicit categorization of **the AI tool's risk level** (low, moderate, high).
- No detailed documentation of **ethical review**.
- The study does not mention long-term **continuous performance monitoring**, and **does not have sufficient follow-up data** to report sensitivity, specificity, false-negative rates, interval cancers or cancer stage at diagnosis.



Reference

- Wells BJ, Nguyen HM, McWilliams A, Pallini M, Bovi A, Kuzma A, Kramer J, Chou SH, Hetherington T, Corn P, Taylor YJ. A practical framework for appropriate implementation and review of artificial intelligence (FAIR-AI) in healthcare. *NPJ digital medicine*. 2025 Aug 11;8(1):514.
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Thank You