

# AI and Cardiology

**Partho P Sengupta, MD, FACC**

Henry Rutgers Professor of Cardiology,  
Chief of Cardiology, Rutgers Robert Wood Johnson Medical  
School & Chief of Cardiac Services, Robert Wood Johnson  
University Hospital, New Brunswick, NJ

Oct 11<sup>th</sup>, 2025



Kentucky  
CHAPTER



# Disclosures

Within the prior 24 months, I have had a financial relationship with a company producing, marketing, selling, re-selling, or distributing healthcare products used by or on patients:

## **Nature of Financial Relationship**

Grant/Research Support

Consultant Fees/Honoraria

Individual Stock(s)/Stock Options

## **Ineligible Company**

HeartSciences, Butterfly Inc., MindMics

RCE technologies, HeartSciences

RCE technologies, HeartSciences

Supported by Grants from NHLBI (1R01HL173998-01A1, 3U01HL088942-17S1, 1P50MD017356-01), & NSF-Bridges in Digital Health



# Democratization of AI

Training compute (FLOPs) of milestone Machine Learning systems over time



<https://arxiv.org/abs/2202.05924>

## Artificial Intelligence

The field of study

## Machine Learning

The branch of AI that focuses on the autonomous learning of tasks

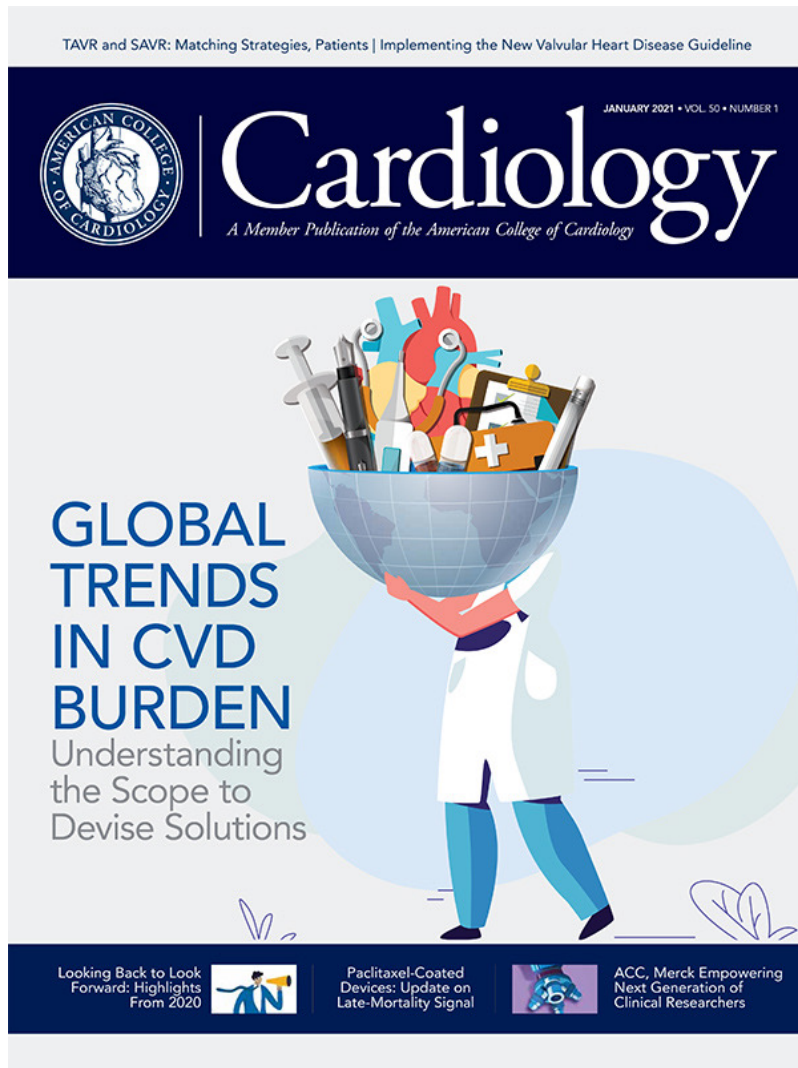
## Deep Learning

The subset of machine learning that uses an artificial neural network architecture to perform complex tasks  
Example: CNN

## Generative AI

Deep learning that creates data similar to its training set  
Example: LLM

# “Will AI Displace Cardiology Workforce?”





# THE LANCET

## Gastroenterology & Hepatology

[This journal](#) [Journals](#) [Publish](#) [Clinical](#) [Global health](#) [Multimedia](#) [Events](#) [About](#)

ARTICLES · [Volume 10, Issue 10](#), P896-903, October 2025

[Download Full Issue](#)

## Endoscopist deskilling risk after exposure to artificial intelligence in colonoscopy: a multicentre, observational study

[Krzysztof Budzyń, MD](#)<sup>a,b</sup> · [Marcin Romańczyk, MD](#) <sup>a,b</sup>  · [Diana Kitala, PhD](#)<sup>c</sup> · [Paweł Kołodziej, MD](#)<sup>d</sup> · [Marek Bugajski, MD](#)<sup>e</sup> · [Hans O Adami, MD](#)<sup>f,g</sup> · et al. [Show more](#)

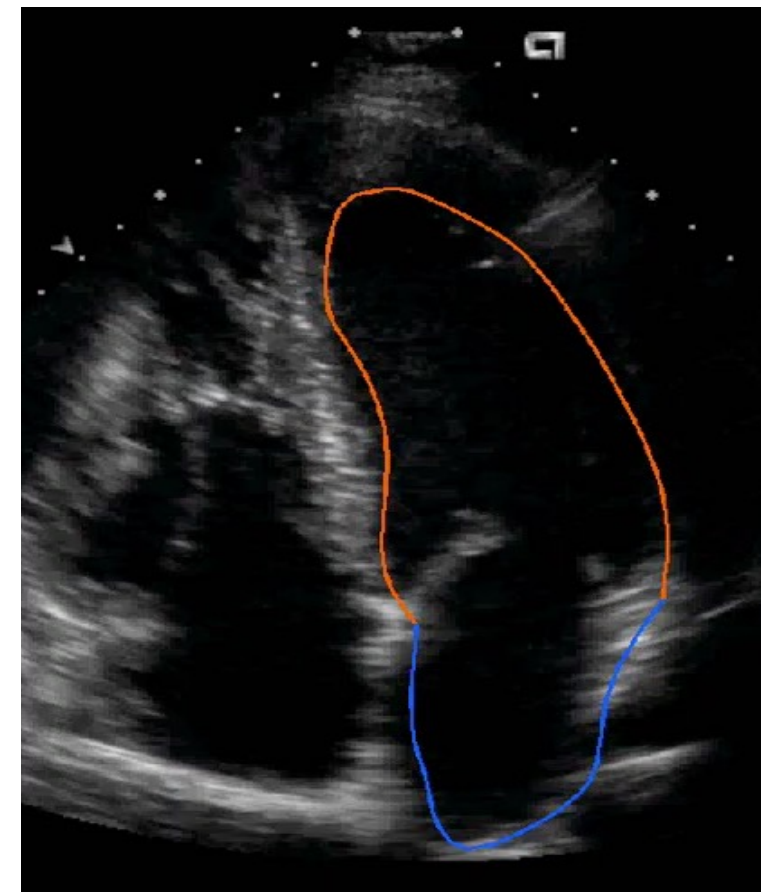
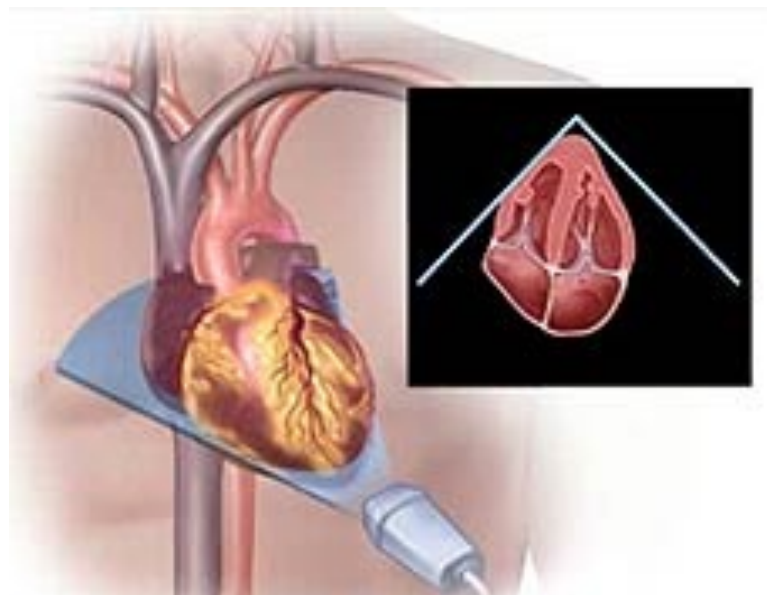
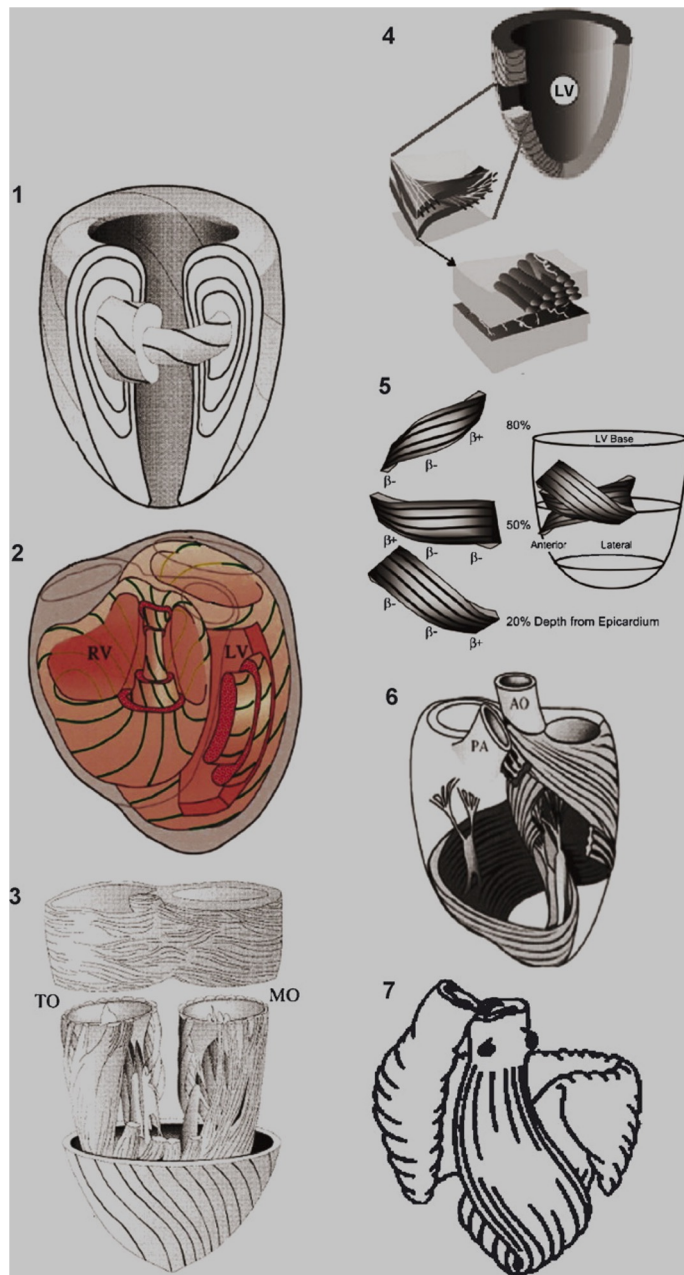
[Affiliations & Notes](#)  [Article Info](#)  [Linked Articles \(3\)](#) 



Until 2007 there  
were 8 different  
models of heart's  
function.

Gilbert et al. Eur J Cardiothorac Surg 2007;32:231-49



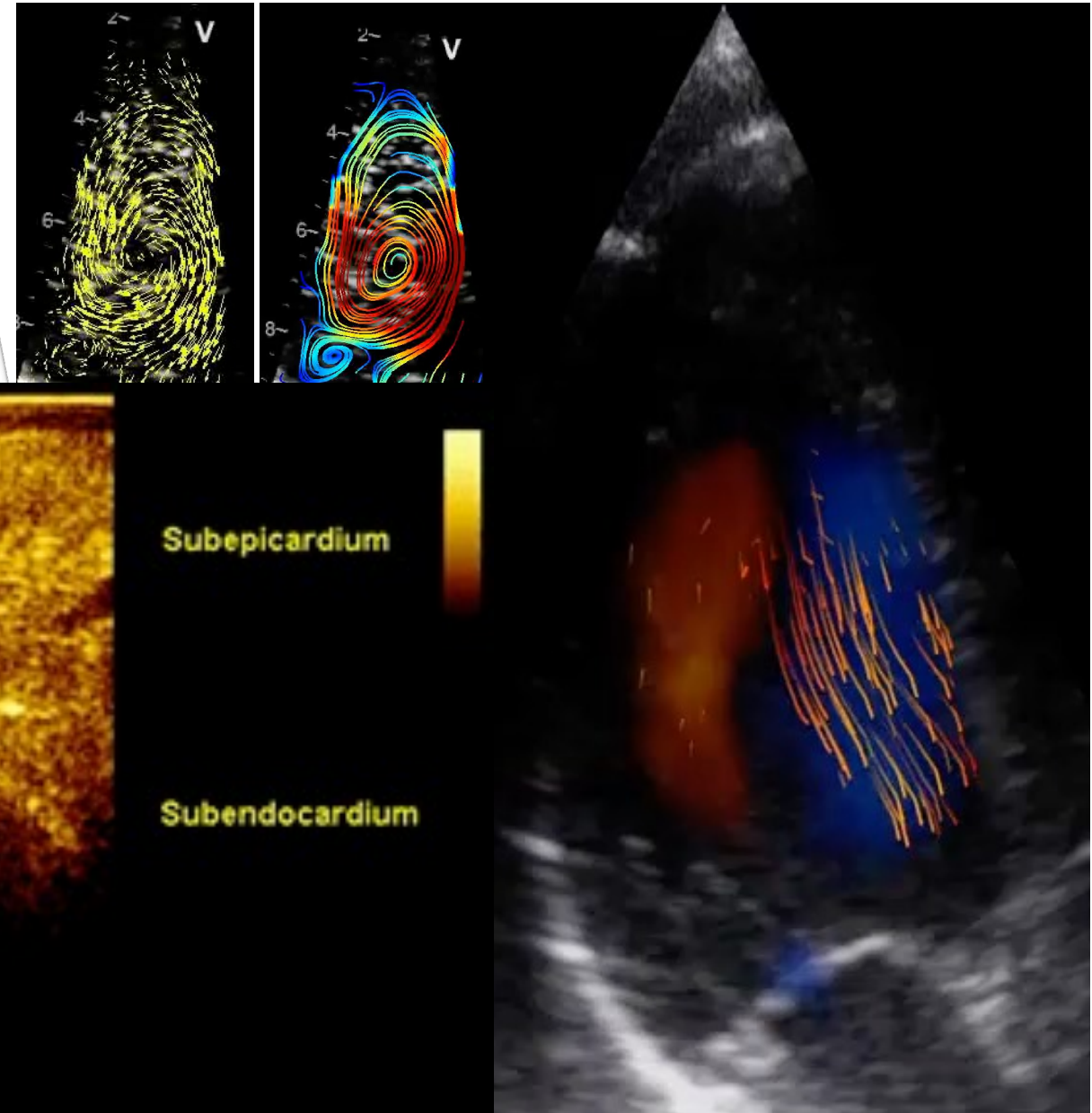


# Left Ventricular Muscle And Fluid Mechanics

## Left Ventricular Structure and Function

**Basic Science for Cardiac Imaging**  
 Partho P. Sengupta, MBBS, MD, DM,\* Josef Korinek, MD,\*  
 Marek Belohlavek, MD, PhD, FACC, FSC,\* Jagat Narula, MBBS, MD, DM, PhD, FACC, FAHA  
 Mani A. Vannan, MBBS, FACC,† Arshed Jahangir, MD, FACC,\*  
 Bijoy K. Khandheria, MD, FESC, FASE, FACC‡  
*Rochester, Minnesota; Irvine, California; and Scottsdale, Arizona*

The myofiber geometry of the left ventricle (LV) changes gradually from a right-handed helix in the subendocardium to a left-handed helix in the subepicardium. In this review, we associate the LV myofiber architecture with emerging concepts of the electromechanical arrangement in a beating heart. We discuss: 1) the morphogenesis and anatomical arrangement of muscle fibers in the adult LV; 2) the sequence of depolarization and repolarization; 3) the sequence in longitudinal and circumferential deformation; 4) the sequence of LV rotation and torsion; 5) the inhomogeneity of transmural myocardial deformation; 6) the sequence of longitudinal and circumferential deformation and the intracavitary flow direction observed during sequences of LV deformation and the link between LV deformation and electrical activation and the link between each phase of the cardiac cycle. Integrating the LV structure with electrical activation and motion sequences observed *in vivo* provides an understanding about the spatiotemporal sequence of regional myocardial performance that is essential for noninvasive cardiac motion sequences observed *in vivo*. (J Am Coll Cardiol 2006;48:1988–2001) © 2006 by the American College of Cardiology Foundation

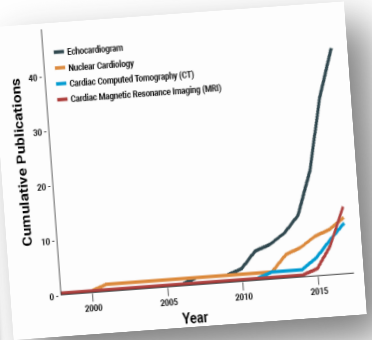
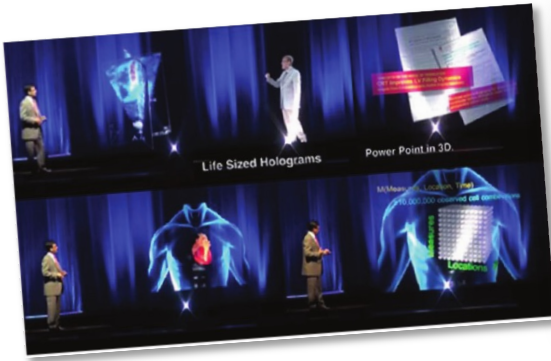






# Intelligent Platforms for Disease Assessment

## Novel Approaches in Functional Echocardiography



### Machine Learning for Data-Driven Discovery

The Rise and Relevance\*

### Enforcing Quality in Strain Imaging Through AI-Powered Surveillance\*

### Automated Interpretation of Myocardial Perfusion Images

Tell Me AI Where to Look\*

### Predicting Preclinical Heart Failure Progression

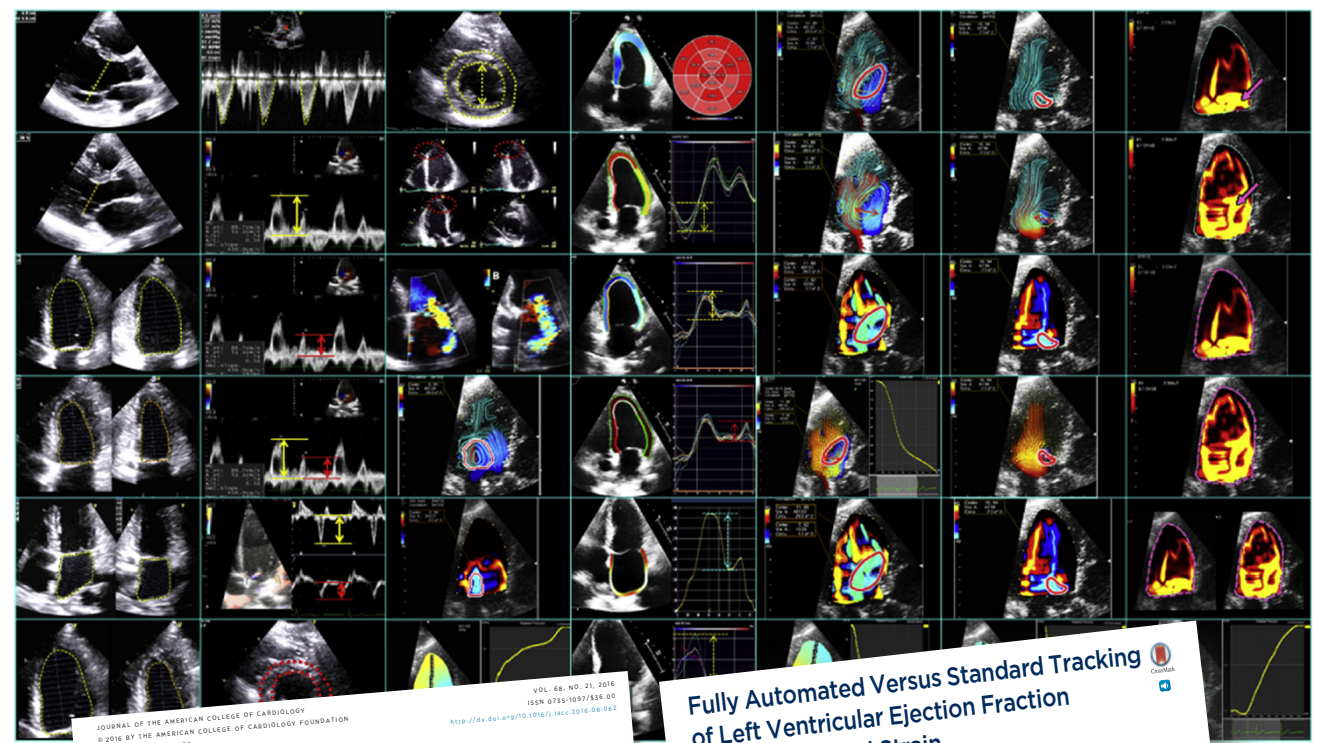
The Rise of Machine-Learning for Population Health\*

### Building Trust in AI

Opportunities and Challenges for Cardiac Imaging

### Imaging With Deep Learning

Sharpening the Cutting Edge



#### Machine-Learning Algorithms to Automate Morphological and Functional Assessments in 2D Echocardiography

Sukrit Narula, BS,\* Khader Shameer, PhD,\* Alaa Mabrouk Salem Omar, MD, PhD,\*<sup>1,2</sup> Joel T. Dudley, PhD,\* Partho P. Sengupta, MD, DM\*

**ABSTRACT**

**BACKGROUND** Machine-learning models may aid cardiac phenotypic recognition by using features of cardiac tissue deformation.

**OBJECTIVES** This study investigated the diagnostic value of a machine-learning framework that incorporates speckle-tracking echocardiographic data for automated discrimination of hypertrophic cardiomyopathy (HCM) from physiological hypertrophy seen in athletes (ATH).

**METHODS** Expert-annotated speckle-tracking echocardiographic datasets obtained from 77 ATH and 62 HCM patients were used for developing an automated system. An ensemble machine-learning model with 3 different machine-learning algorithms (support vector machines, random forests, and artificial neural networks) was developed and a majority voting method was used for conclusive predictions with further K-fold cross-validation.

**RESULTS** Feature selection using an information gain (IG) algorithm revealed that volume was the best predictor for differentiating between HCM and ATH (IG = 0.24) followed by mid-left ventricular segmental (IG = 0.134) and average longitudinal strain (IG = 0.131). The ensemble machine-learning model showed increased sensitivity and specificity compared with early-to-late diastolic transmitral velocity ratio ( $p < 0.01$ ), average early diastolic tissue velocity ( $e'$ ) ( $p < 0.01$ ), and strain ( $p = 0.04$ ). Because ATH were younger, adjusted analysis was undertaken in younger HCM patients and compared with ATH with left ventricular wall thickness  $>13$  mm. In this subgroup analysis, the automated model continued to show equal sensitivity, but increased specificity relative to early-to-late diastolic transmitral velocity ratio,  $e'$ , and strain.

**CONCLUSIONS** Our results suggested that machine-learning algorithms can assist in the discrimination of physiological versus pathological patterns of hypertrophic remodeling. This effort represents a step toward the development of a real-time, machine-learning-based system for automated interpretation of echocardiographic images, which may help novice readers with limited experience. (J Am Coll Cardiol 2016;68:2287-95) © 2016 by the American College of Cardiology Foundation.

#### Fully Automated Versus Standard Tracking of Left Ventricular Ejection Fraction and Longitudinal Strain

The FAST-EFs Multicenter Study

Christian Knackstedt, MD,\* Sebastian C.A.M. Bekkers, MD, PhD,\* Georg Schummers, MD, PhD,\* Marcus Schreckenberg, MD, PhD,\* Denisa Muraru, MD, PhD,\* Luigi P. Badano, MD, PhD,\* Andreas Franke, MD,\* Chirag Bavishi, MD, MPH,\* Alaa Mabrouk Salem Omar, MD, PhD,\* Partho P. Sengupta, MD, DM

**ABSTRACT**

**BACKGROUND** Echocardiographic determination of ejection fraction (EF) by manual tracing of endocardial borders is time consuming and operator dependent, whereas visual assessment is inherently subjective.

**OBJECTIVES** This study tested the hypothesis that a novel, fully automated software using machine learning-enabled image analysis will provide rapid, reproducible measurements of left ventricular volumes and EF, as well as average biplane longitudinal strain (LS).

**METHODS** For a total of 255 patients in sinus rhythm, apical 4- and 2-chamber views were collected from 4 centers that assessed EF using both visual estimation and manual tracing (biplane Simpson's method). In addition, datasets were saved in a centralized database, and machine learning-enabled software (AutoLV, TomTec-Arena 1.2, TomTec Imaging Systems, Unterschleißheim, Germany) was applied for fully automated EF and LS measurements. A reference center reanalyzed all datasets (by visual estimation and manual tracing), along with manual LS determinations.

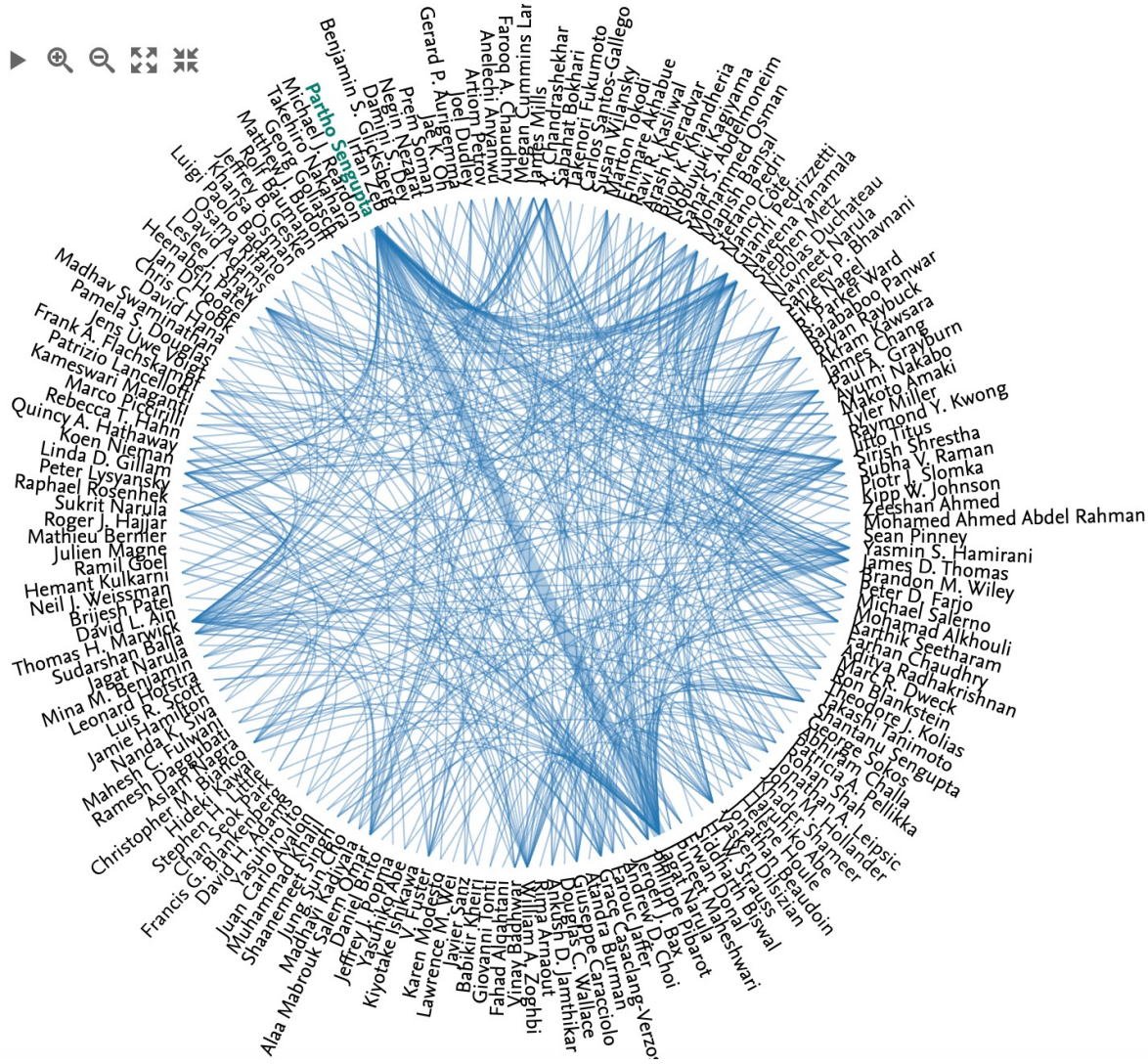
**RESULTS** AutoLV measurements were feasible in 98% of studies, and the average analysis time was  $8 \pm 1$  s/patient. Interclass correlation coefficients and Bland-Altman analysis revealed good agreements among automated EF, local center manual tracking, and reference center manual tracking, but not for visual EF assessments. Similarly, automated and manual LS measurements obtained at the reference center showed good agreement. Intraobserver variability was higher for visual EF than for manual EF or manual LS, whereas interobserver variability was higher for both visual and manual EF, but not different for LS. Automated EF and LS had no variability.

**CONCLUSIONS** Fully automated analysis of echocardiography images provides rapid and reproducible assessment of left ventricular EF and LS. (J Am Coll Cardiol 2015;66:1456-66) © 2015 by the American College of Cardiology Foundation.



# Automated Machine Learning

Circle





Circulation

EDITORIAL

# Learning About Machine Learning to Create a Self-Driving Echocardiographic Laboratory

## Technical Considerations

Article, see p 1623

Jan D'hooge, PhD  
Alan G. Fraser, BSc, MBChB

Circulation

EDITORIAL

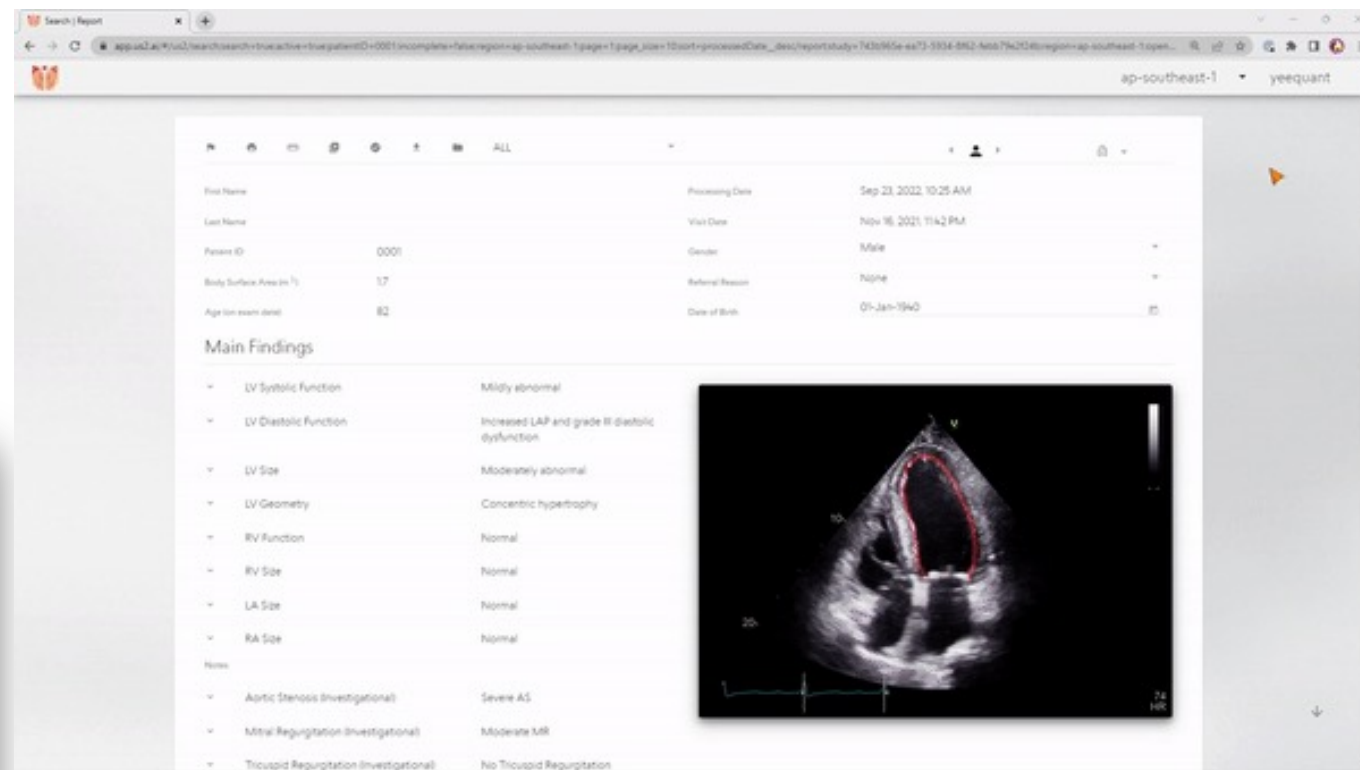
# Will Artificial Intelligence Replace the Human Echocardiographer?

## Clinical Considerations

Article, see p XXX

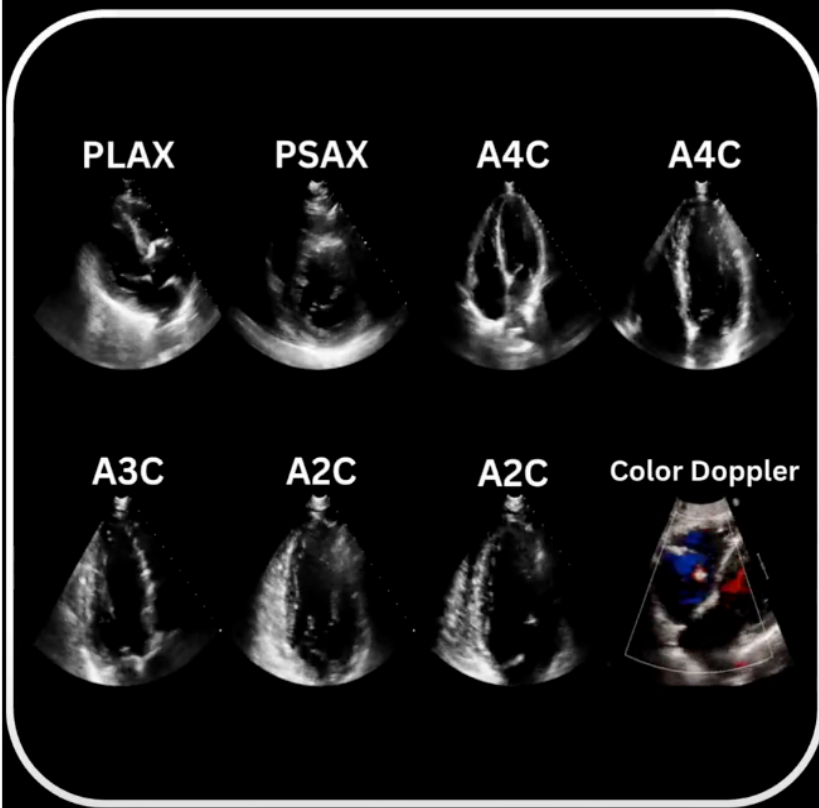
Partho P. Sengupta, MD,  
DM  
Donald A. Adjero, PhD

**A**n older population with an increased prevalence of cardiovascular disease and an aging workforce are engendering a state of healthcare crisis in cardiology.<sup>1</sup> Most cardiologists now face an unprecedented time crunch as they



US2AI, courtesy Yoram Hummel

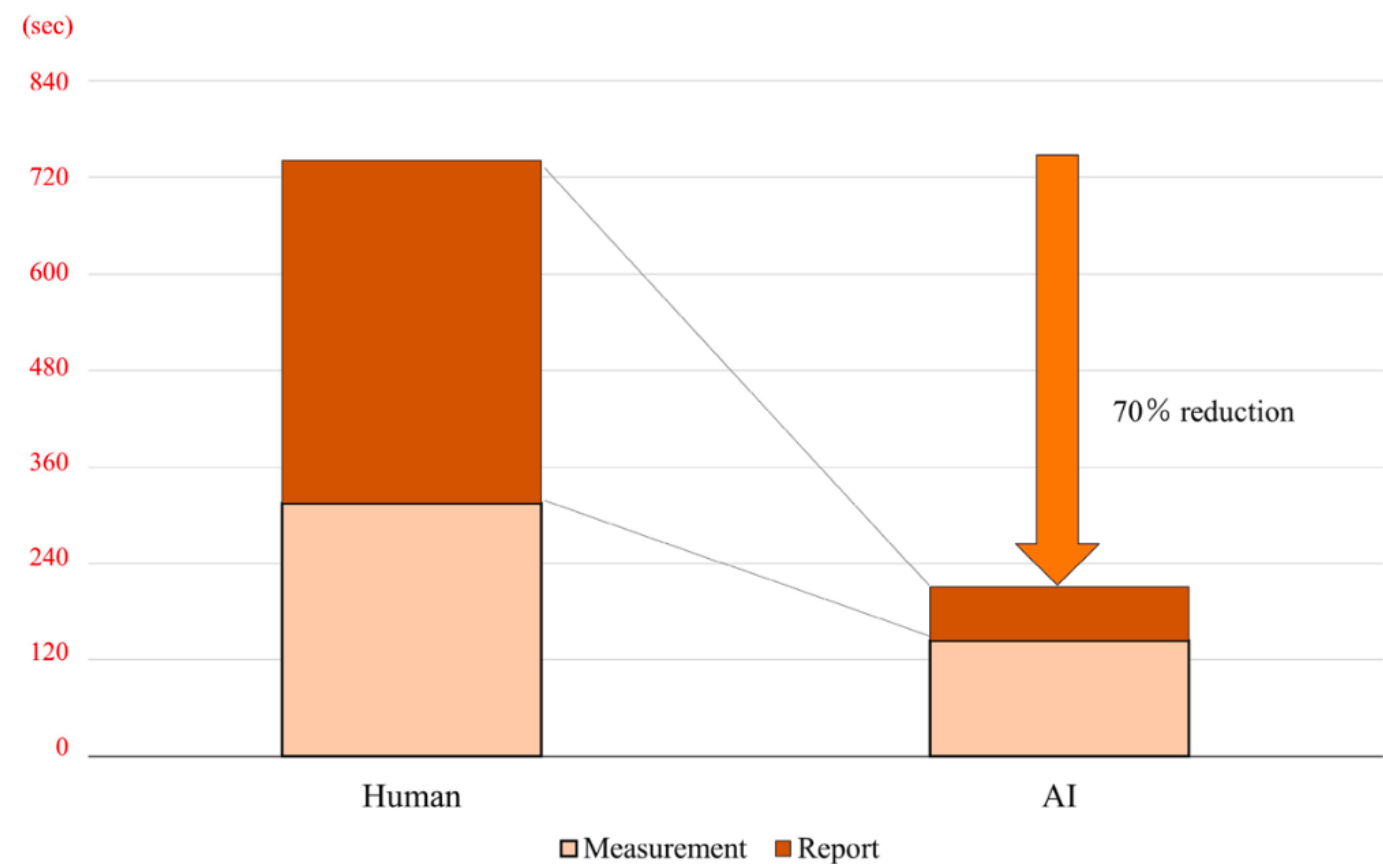
## Multi-view Echocardiography



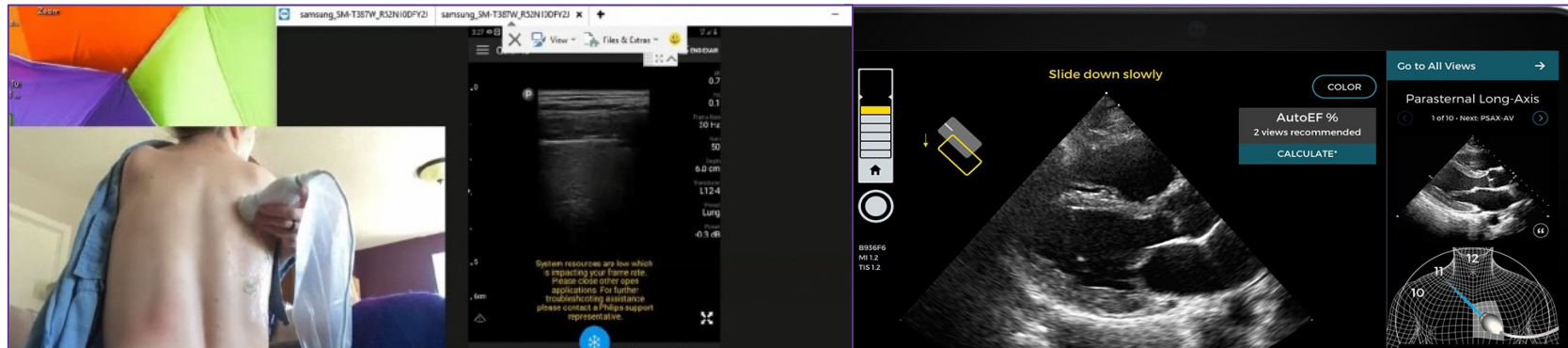
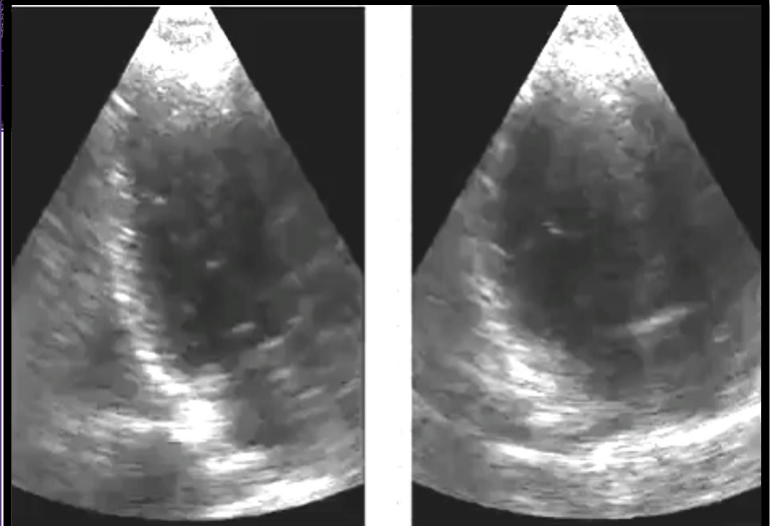


# Reducing echocardiographic examination time through routine use of fully automated software: a comparative study of measurement and report creation time

Yukina Hirata<sup>1</sup> · Yuka Nomura<sup>1</sup> · Yoshihito Saijo<sup>2</sup> · Masataka Sata<sup>2</sup> · Kenya Kusunose<sup>2,3</sup> 



# AI-driven Autonomous Imaging, Miniaturized/ Portable Devices

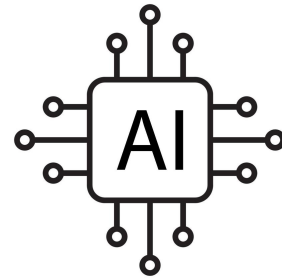
| Self Scanning                                                                      |                                                                                                                                                                                                                                                                                                                                                                                                     | Wearable Ultrasound                                                                  |  |
|------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|--------------------------------------------------------------------------------------|--|
|  |                                                                                                                                                                                                                                                                                                                                                                                                     |   |  |
| Portable CT                                                                        | <div><div></div><div></div><div></div><div></div></div> |  |  |
| Portable CT                                                                        | Portable MRI                                                                                                                                                                                                                                                                                                                                                                                        | Fully Automated Imaging                                                              |  |



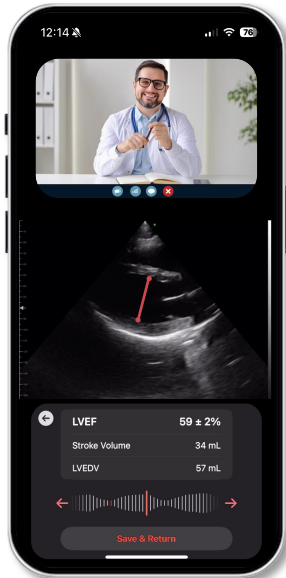
# Autonomous Ultrasound Scanner



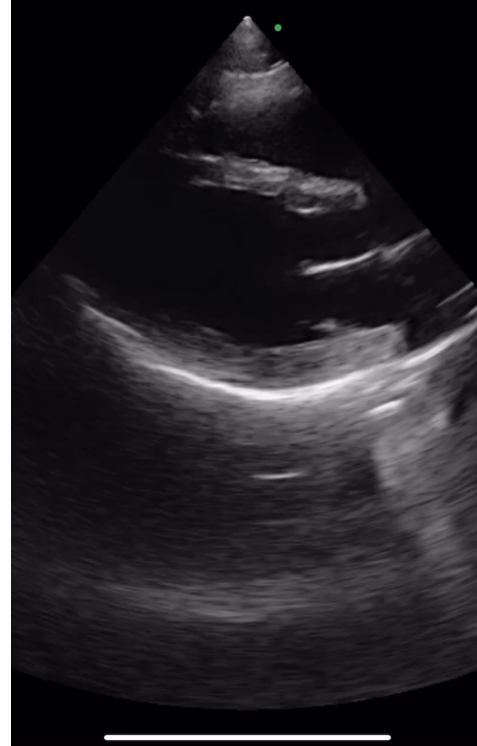
**Hands-free**



**AI-powered**



Ultrasound System



 Easy setup with nurses or family members

 Simplified workflow with reduced clinical overhead

 Access anytime, anywhere

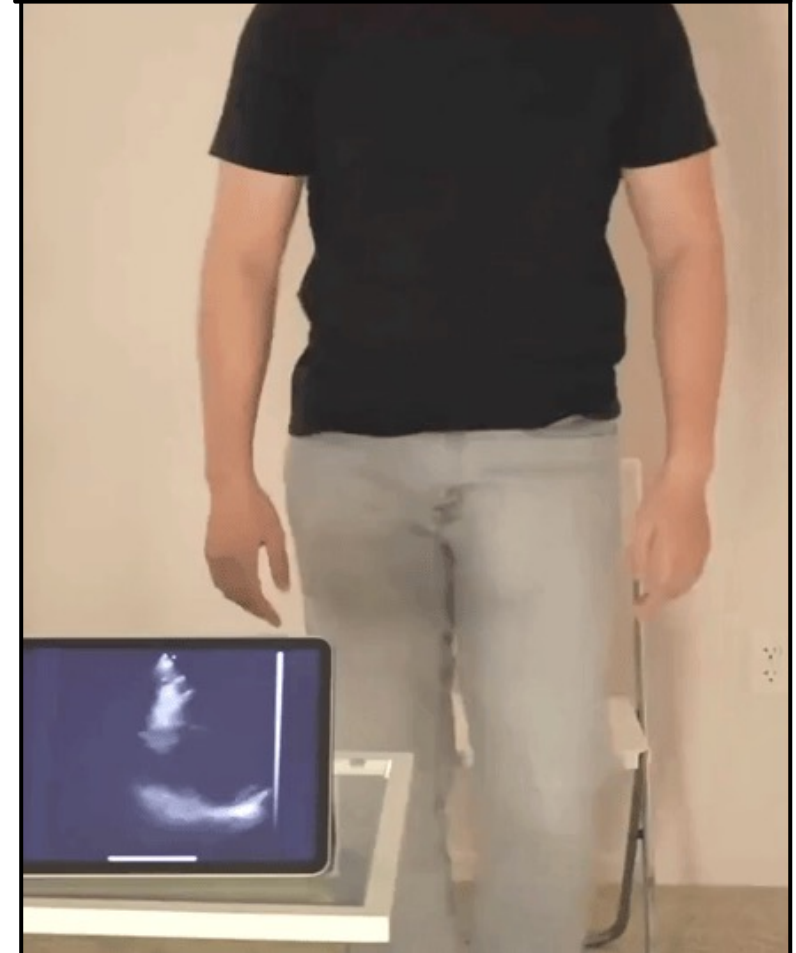
**nature**

[Explore content](#) ▾ [About the journal](#) ▾ [Publish with us](#) ▾

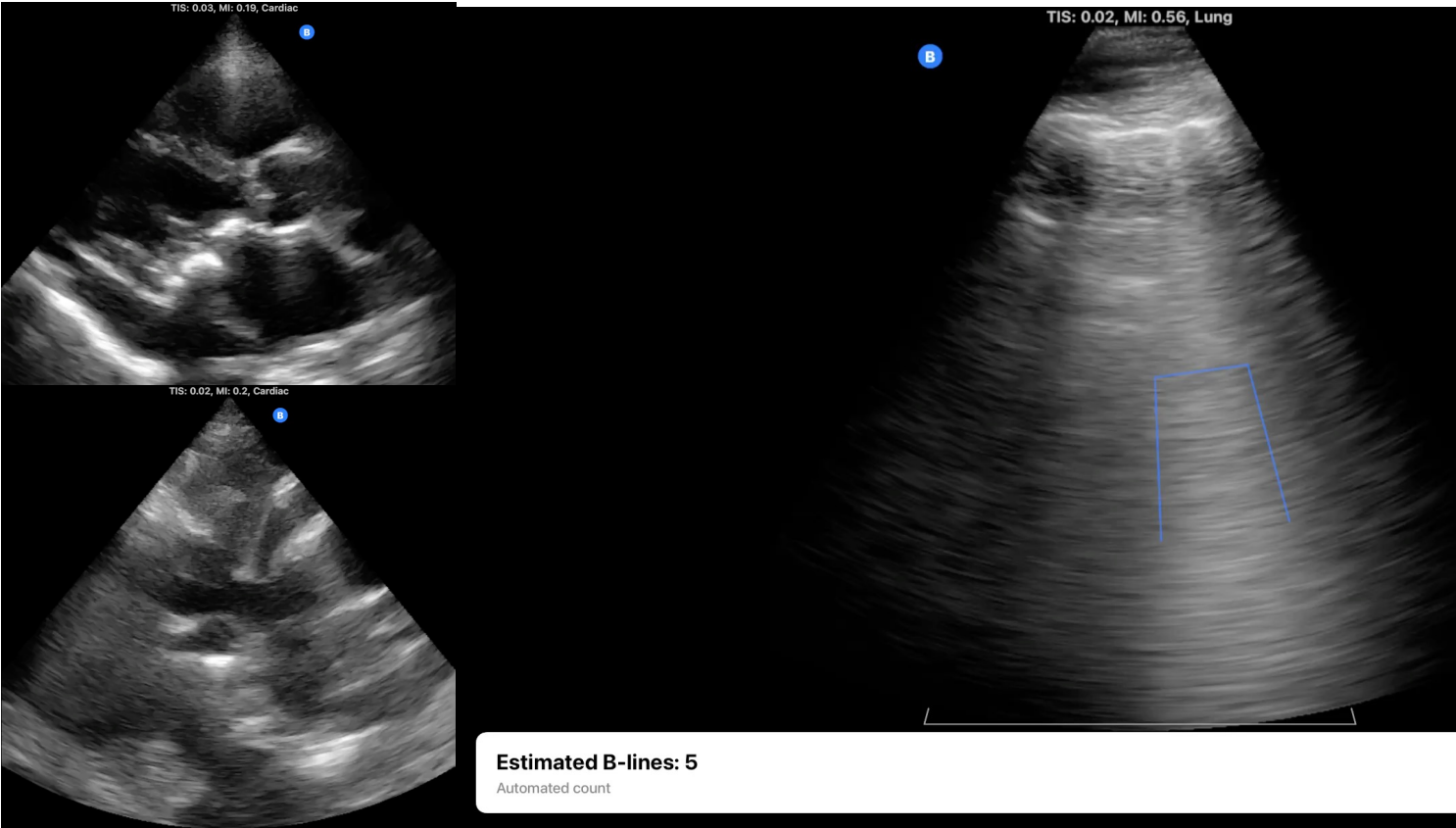
[nature](#) > [articles](#) > [article](#)

Article | [Open access](#) | Published: 25 January 2023

## A wearable cardiac ultrasound imager



# POCUS-CARE Trial: A Step-Wedge Cluster Randomized Trial



| Outcome                                                                     | POCUS     | Control   | p-value |
|-----------------------------------------------------------------------------|-----------|-----------|---------|
| Cumulative LOS in each arm (days)                                           | 713       | 959       | 0.008** |
| Cumulative total direct hospitalization cost for each arm (\$ all patients) | 2,442,945 | 3,194,482 | 0.020** |



# The Need

## Projected Burden of CV Disease and Workforce Shortage



**35.6 million** CV deaths by 2050



Total CV deaths **↑ 73.4%**

Age-standardised CV mortality rates **↓ 30.5%**



European Journal of Preventive Cardiology (2024) 00, 1–15  
<https://doi.org/10.1093/eurjpc/zwae281>

**FULL RESEARCH PAPER**  
Cardiovascular disease

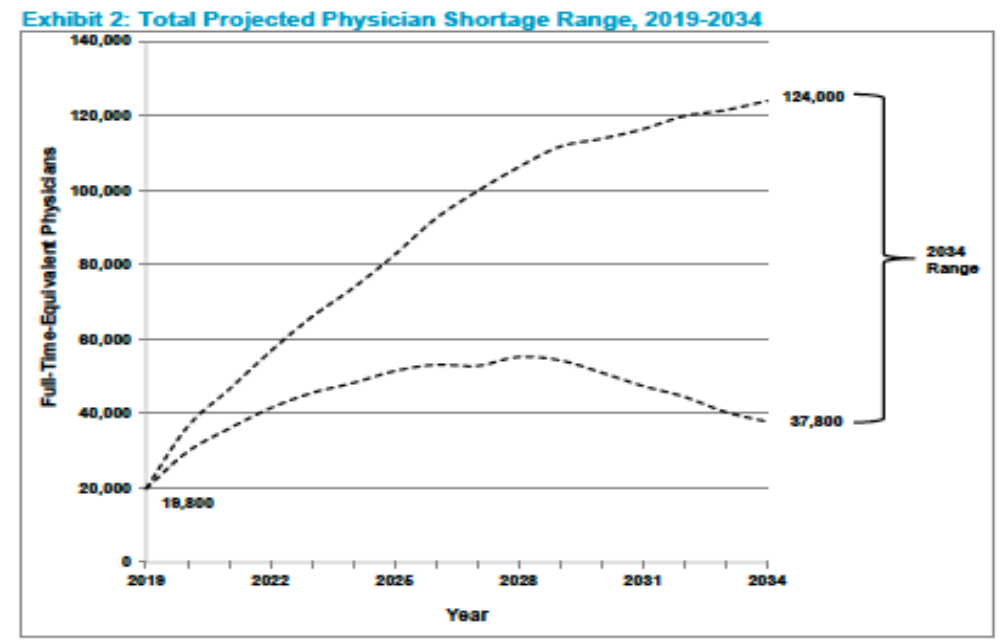
**Global burden of cardiovascular diseases:  
projections from 2025 to 2050**

**10 million  
more health workers**

are needed by 2030, primarily in low and lower middle income countries

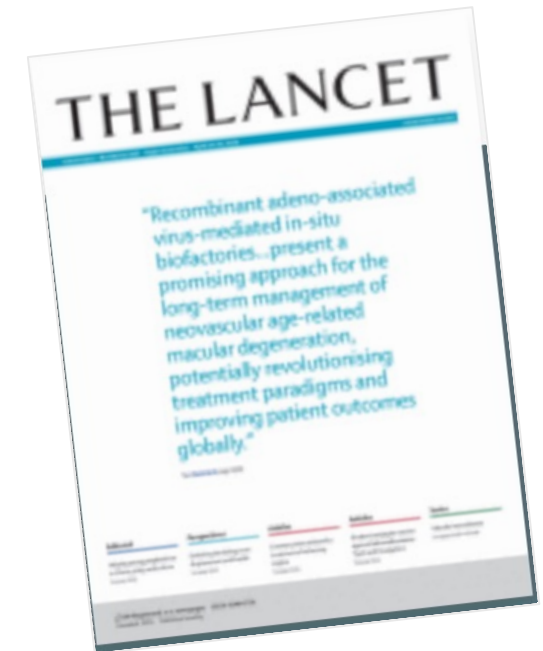
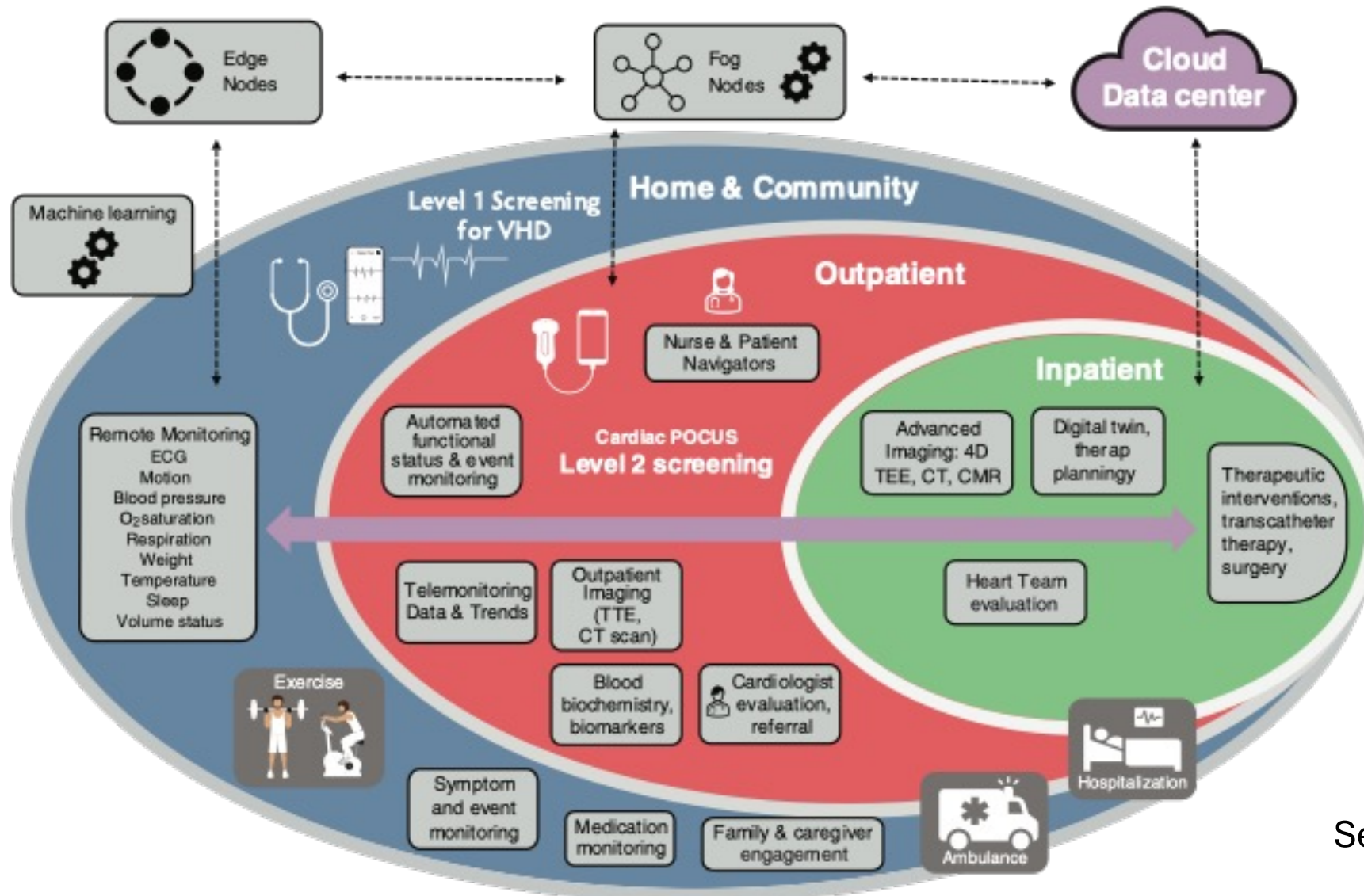
The Complexities of Physician Supply and Demand:  
Projections From 2019 to 2034

AAMC



# The Need

## AI Can Advance Access For Screening and Early Detection of Heart Disease

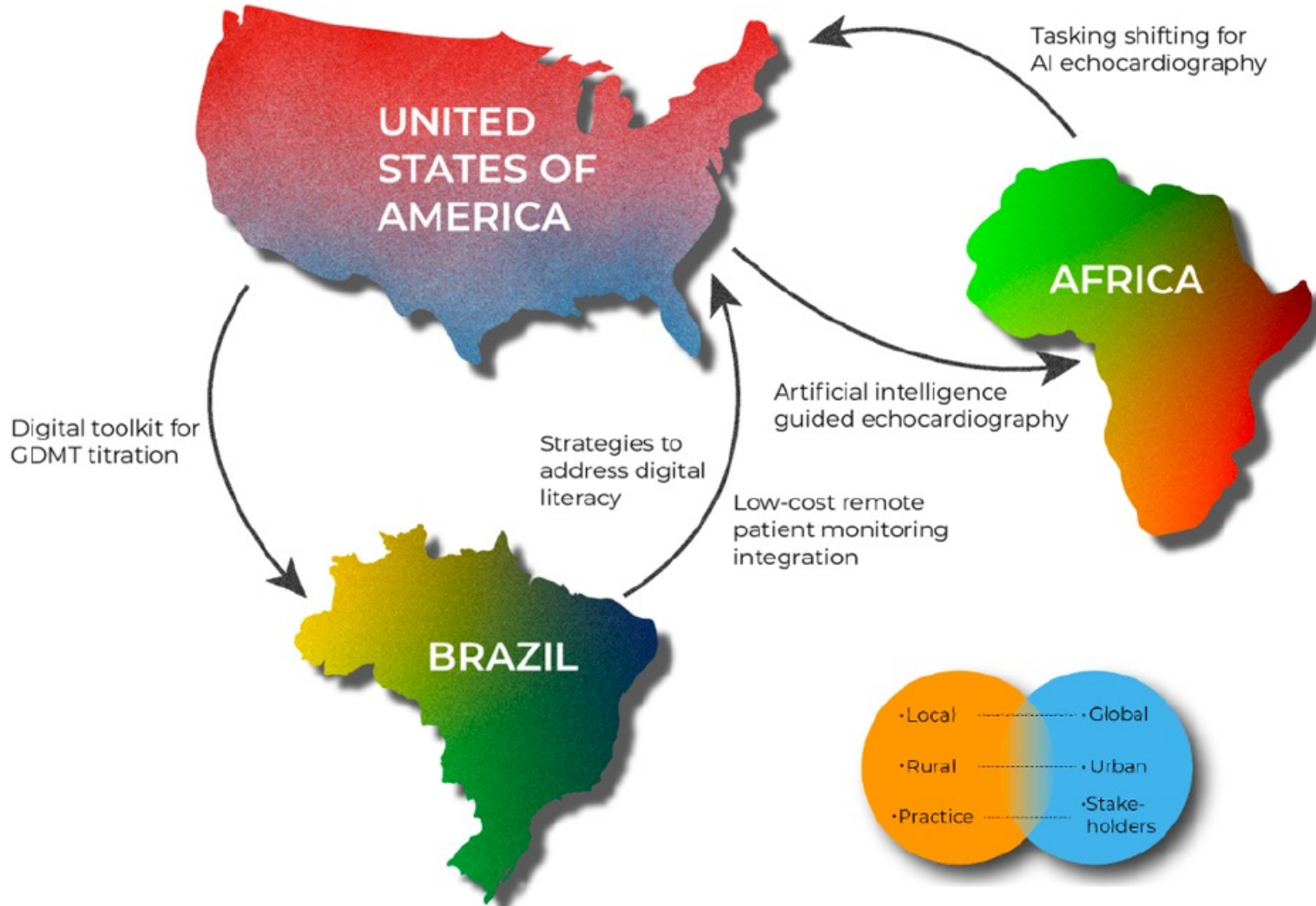


Sengupta PP et al, Lancet 2023



# The Need

## AI Can Advance Global Cardiovascular Health Equity



Medical imaging clinic in Juana Vicente, Dominican Republic. US probe (left) is connected to a laptop computer, powered by an external battery (orange cord) because of absent electricity.



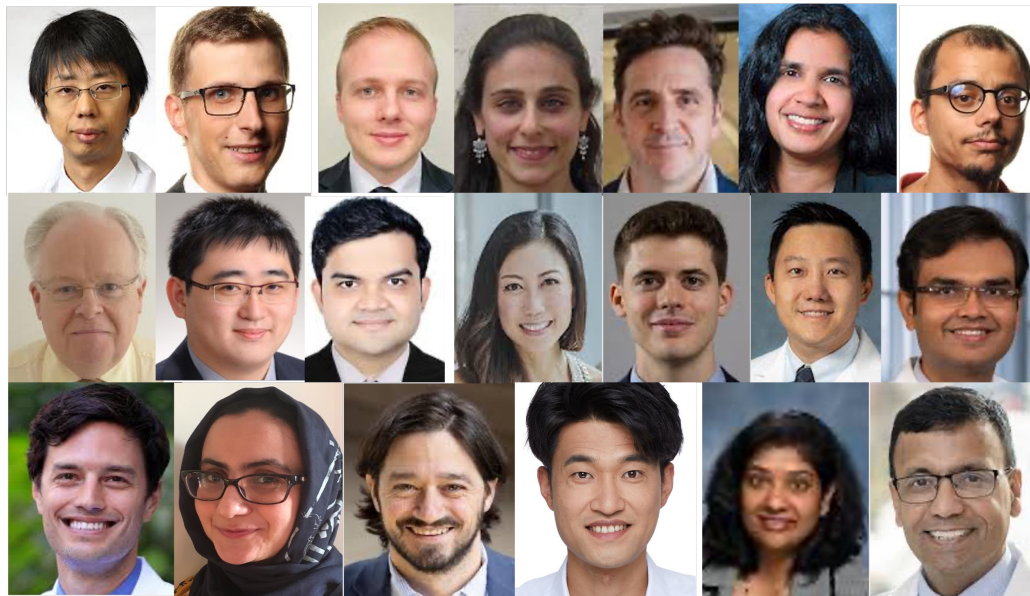
**Prediction # 1.** A new era of "Augmented Imaging"—physicians using AI to boost efficiency, reduce burnout, and rediscover joy in medicine, enhancing both clinical care and their own professional fulfillment.

[JACC Journals](#) › [JACC: Imaging](#) › [Archives](#) › [Just Accepted](#)

 **GET ACCESS** | State-of-the-Art Review | 27 August 2025



## PRIME 2.0: An Update to The Proposed Requirements for Cardiovascular Imaging-Related Machine Learning Evaluation Checklist

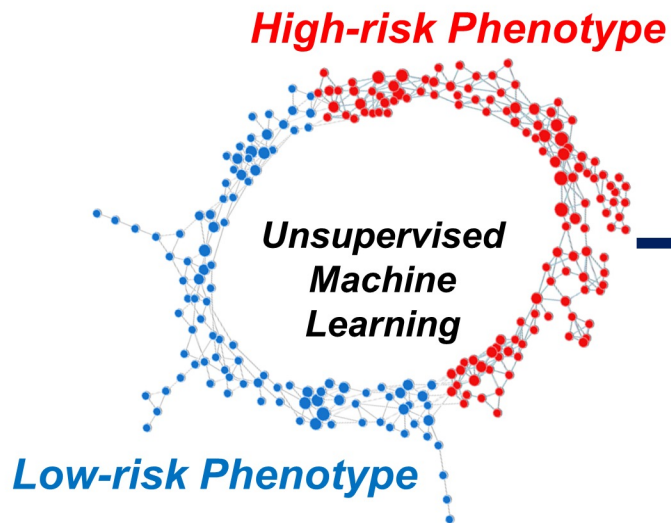




# Deep Learning for Assessing Diastolic Dysfunction in HFpEF

Ambarish Pandey, MD

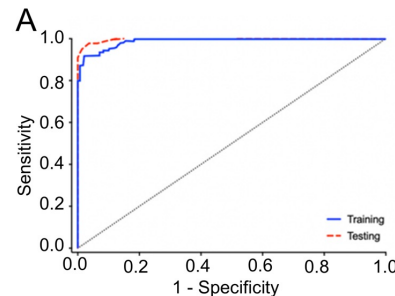
## Similarity Clustering



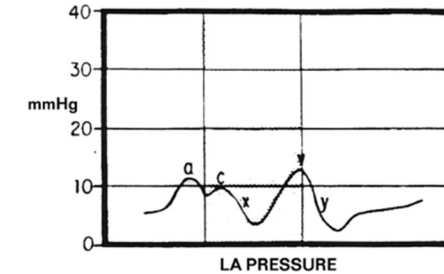
## Supervised Classifier



## Deep Learning Model

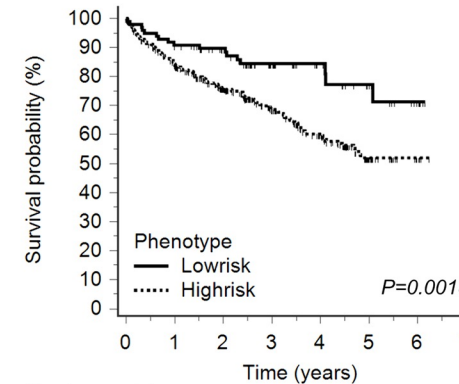


## Model Evaluation



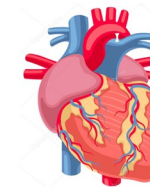
### Hemodynamic Validation

Invasive cohort  
(N = 83)



### Outcome Validation

- TOPCAT (N = 518)
- External Validation (N = 219)



Cardiac Biomarkers



Exercise Performance

- NEAT (N = 110)
- RELAX (N = 216)



**High-risk Phenotype**

**Unsupervised  
Machine  
Learning**

**Low-risk Phenotype**

**J Am Coll Cardiol Img 2021, 14: 1887–1900**

**A Deep Learning Classifier for Assessing Diastolic Dysfunction Severity in HFrEF** Logout

Enter the values

|                                                   |                                               |
|---------------------------------------------------|-----------------------------------------------|
| Septal Early Diastolic Relaxation Velocity (cm/s) | LV Mass Indexed to BSA (g/m <sup>2</sup> )    |
| <input type="text" value="e_prime"/>              | <input type="text" value="LVMI"/>             |
| Late Diastolic Transmittal Velocity (cm/s)        | LA Volume Indexed to BSA (ml/m <sup>2</sup> ) |
| <input type="text" value="A"/>                    | <input type="text" value="LAVI"/>             |
| E/e' Ratio (ratio)                                | LV Ejection Fraction (%)                      |
| <input type="text" value="E_e_prime"/>            | <input type="text" value="EF"/>               |
| Early Diastolic Peak Mitral Flow (m/s)            | E/A Ratio (ratio)                             |
| <input type="text" value="E"/>                    | <input type="text" value="E_A"/>              |
| Tricuspid Regurgitation Velocity (m/s)            |                                               |
| <input type="text" value="TRV"/>                  |                                               |

Submit

<https://wvu-model.herokuapp.com>

#### GUIDELINES AND STANDARDS

### Recommendations for the Evaluation of Left Ventricular Diastolic Function by Echocardiography and for Heart Failure With Preserved Ejection Fraction Diagnosis: An Update From the American Society of Echocardiography

Sherif F. Nagueh, MD, FASE (Chair), Danita Y. Sanborn, MD, FASE (Co-Chair), Jae K. Oh, MD, FASE, Bonita Anderson, MAppSc, DMU, ACS, FASE, FASA, Kristen Billick, BS, ACS, RCS, RDSC, FASE, Genevieve Derumeaux, MD, PhD, Allan Klein, MD, FASE, Konstantinos Koulogiannis, MD, FASE, Carol Mitchell, PhD, ACS, RDMS, RDSC, RVT, RT(R), FASE, Amil Shah, MD, Kavita Sharma, MD, Otto A. Smiseth, MD, PhD, Honorary FASE, and Teresa S. M. Tsang, MD, FASE, *Houston and Dallas, Texas; Boston, Massachusetts; Rochester, Minnesota; Brisbane, Australia; San Diego, California; Creteil, France; Cleveland, Ohio; Morristown, New Jersey; Madison, Wisconsin; Baltimore, Maryland; Oslo, Norway; and Vancouver, British Columbia, Canada*

“Machine learning models based on clinical outcomes offer a more evidence-based approach to optimizing the classification of diastolic function with both reduction of “indeterminate” cases and improvement of the clinical relevance of diastolic function classification.”

# Prediction #3 A New Taxonomy of Measurements

## Ultrasonic Texture Features for Assessing Cardiac Remodeling and Dysfunction

Quincy A. Hathaway, PhD,<sup>a,\*</sup> Naveena Yanamala, PhD,<sup>b,\*</sup> Nanda K. Siva, BS,<sup>a</sup> Donald Adjero, PhD,<sup>c</sup> John M. Hollander, PhD,<sup>d</sup> Partho P. Sengupta, MD, DM<sup>b</sup>

### ABSTRACT

**BACKGROUND** Changes in cardiac size, myocardial mass, cardiomyocyte appearance, and, ultimately, the function of the entire organ are interrelated features of cardiac remodeling that profoundly affect patient outcomes.

**OBJECTIVES** This study proposes that the application of radiomics for extracting cardiac ultrasonic textural features (ultrasomics) can aid rapid, automated assessment of left ventricular (LV) structure and function without requiring manual measurements.

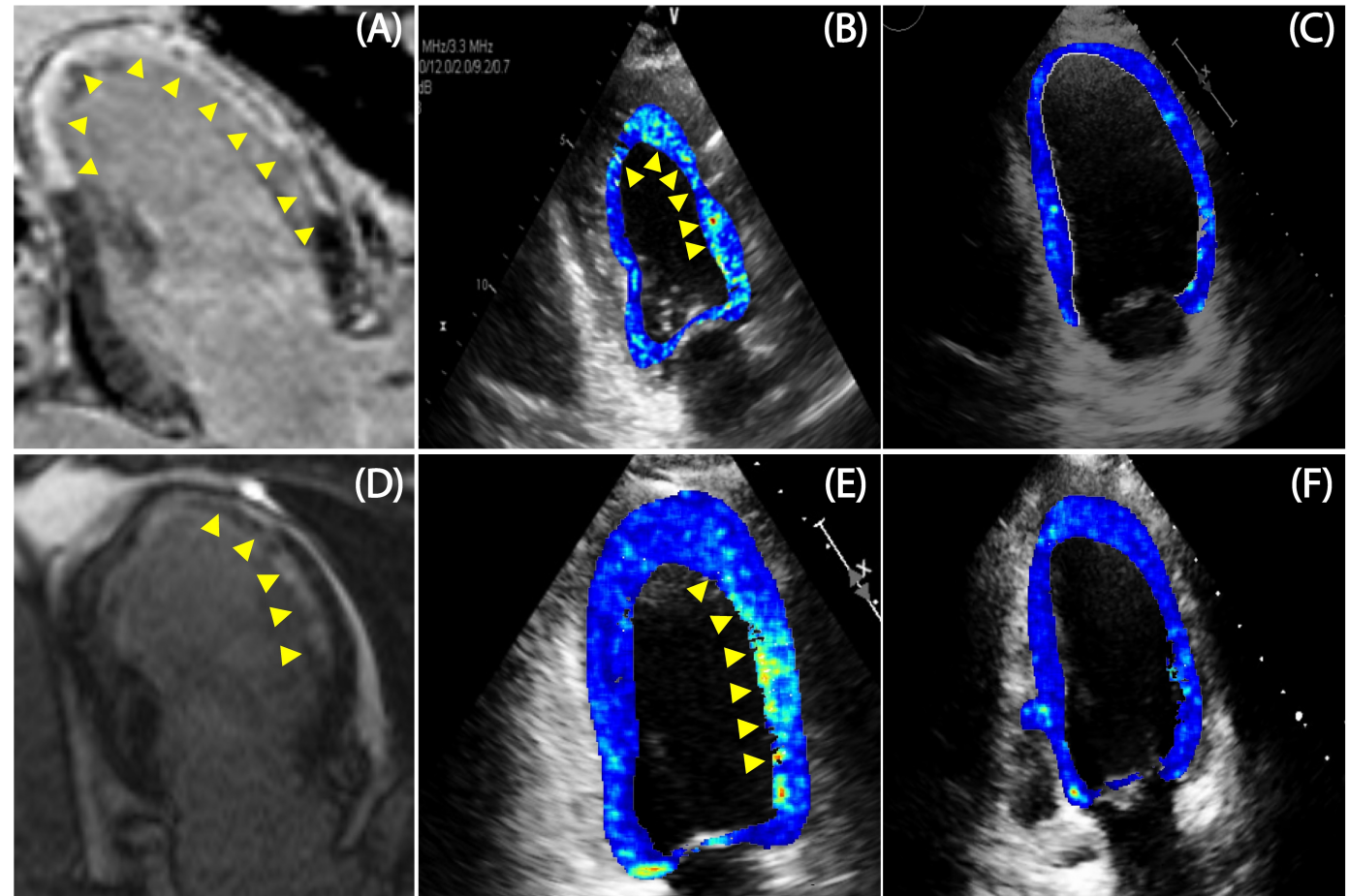
**METHODS** This study developed machine-learning models using cardiac ultrasound images from 1,915 subjects in 3 clinical cohorts: 1) an expert-annotated cardiac point-of-care-ultrasound (POCUS) registry (n = 943, 80% training/testing and 20% internal validation); 2) a prospective POCUS cohort for external validation (n = 275); and 3) a prospective external validation on high-end ultrasound systems (n = 484). In a type 2 diabetes murine model, echocardiography of wild-type (n = 10) and *Lepr<sup>-/-</sup>* (n = 8) mice were assessed longitudinally at 3 and 25 weeks, and ultrasomics features were correlated with histopathological features of hypertrophy.

**RESULTS** The ultrasomics model predicted LV remodeling in the POCUS and high-end ultrasound external validation studies (area under the curve: 0.78 [95% CI: 0.68-0.88] and 0.79 [95% CI: 0.73-0.86], respectively). Similarly, the ultrasomics model predicted LV remodeling was significantly associated with major adverse cardiovascular events in both cohorts ( $P < 0.0001$  and  $P = 0.0008$ , respectively). Moreover, on multivariate analysis, the ultrasomics probability score was an independent echocardiographic predictor of major adverse cardiovascular events in the high-end ultrasound cohort (HR: 8.53; 95% CI: 4.75-32.1;  $P = 0.0003$ ). In the murine model, cardiomyocyte hypertrophy positively correlated with 2 ultrasomics biomarkers ( $R^2 = 0.57$  and  $0.52$ ,  $Q < 0.05$ ).

**CONCLUSIONS** Cardiac ultrasomics-based biomarkers may aid development of machine-learning models that provide an expert-level assessment of LV structure and function. (J Am Coll Cardiol 2022;80:2187-2201) © 2022 by the American College of Cardiology Foundation.



## Ultrasomic Texture Analysis for Predicting Acute MI



Jamthikar A, Sengupta PP et al JACC Cardiovascular Imaging 2025



# Ultrasonic Texture Features for Assessing Cardiac Remodeling and Dysfunction

Quincy A. Hathaway, PhD,<sup>a,\*</sup> Naveena Yanamala, PhD,<sup>b,\*</sup> Nanda K. Siva, BS,<sup>a</sup> Donald Adjero, PhD,<sup>c</sup> John M. Hollander, PhD,<sup>d</sup> Partho P. Sengupta, MD, DM<sup>b</sup>

## ABSTRACT

**BACKGROUND** Changes in cardiac size, myocardial mass, cardiomyocyte ap the entire organ are interrelated features of cardiac remodeling that profound

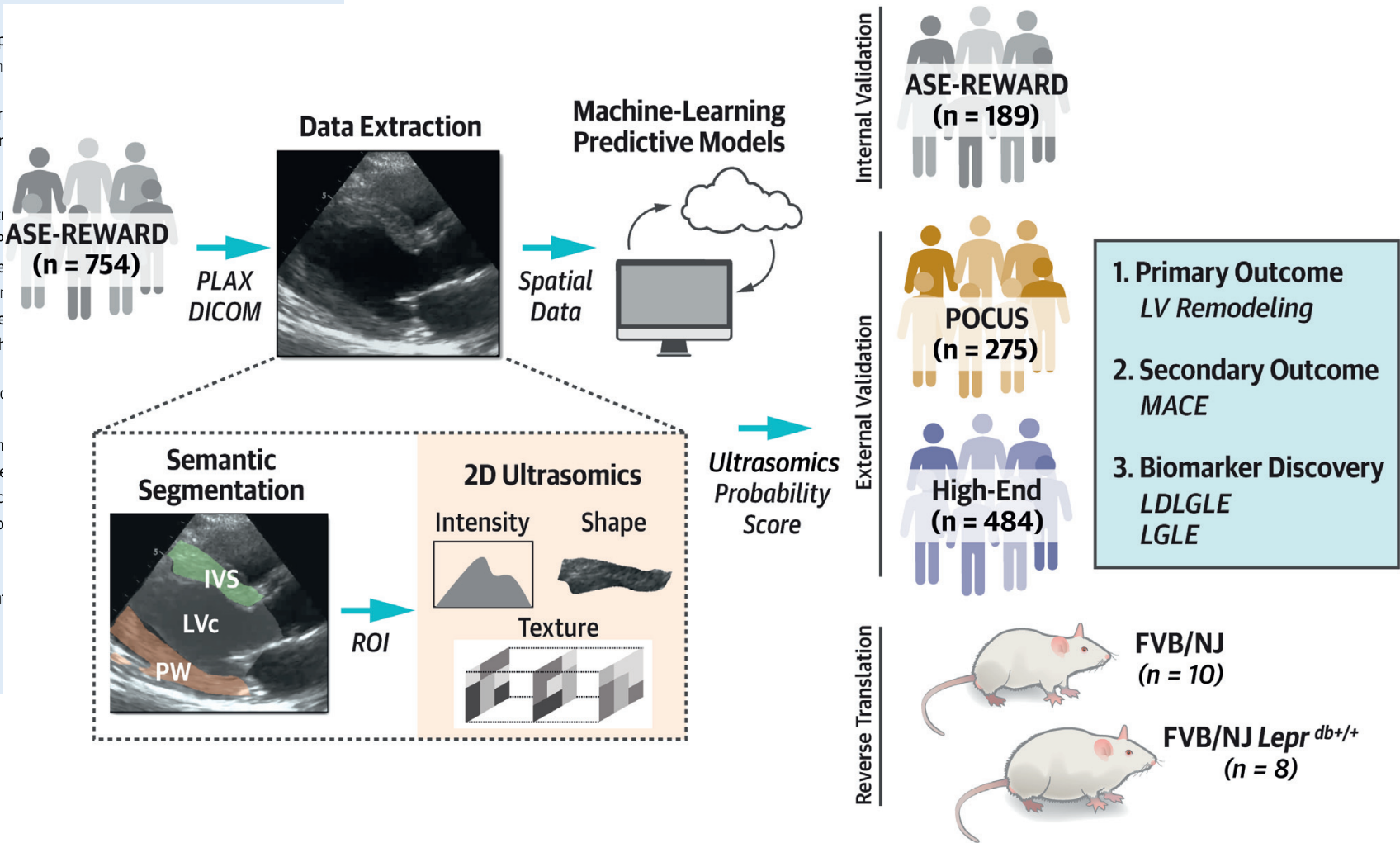
**OBJECTIVES** This study proposes that the application of radiomics for extr (ultrasomics) can aid rapid, automated assessment of left ventricular (LV) str manual measurements.

**METHODS** This study developed machine-learning models using cardiac ult 3 clinical cohorts: 1) an expert-annotated cardiac point-of-care-ultrasound (P testing and 20% internal validation); 2) a prospective POCUS cohort for exte prospective external validation on high-end ultrasound systems (n = 484). In echocardiography of wild-type (n = 10) and Lepr<sup>-/-</sup> (n = 8) mice were asse and ultrasomics features were correlated with histopathological features of t

**RESULTS** The ultrasomics model predicted LV remodeling in the POCUS and studies (area under the curve: 0.78 [95% CI: 0.68-0.88] and 0.79 [95% CI: ultrasomics model predicted LV remodeling was significantly associated with n cohorts ( $P < 0.0001$  and  $P = 0.0008$ , respectively). Moreover, on multivariate was an independent echocardiographic predictor of major adverse cardiovasc cohort (HR: 8.53; 95% CI: 4.75-32.1;  $P = 0.0003$ ). In the murine model, cardio with 2 ultrasomics biomarkers ( $R^2 = 0.57$  and  $0.52$ ,  $Q < 0.05$ ).

**CONCLUSIONS** Cardiac ultrasomics-based biomarkers may aid developmen an expert-level assessment of LV structure and function. (J Am Coll Cardiol American College of Cardiology Foundation.

## CENTRAL ILLUSTRATION Ultrasonics Biomarkers for Predicting Cardiac Outcomes

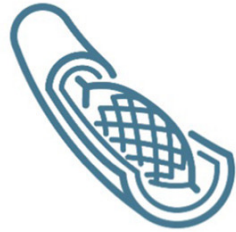




## Coronary Intervention

## Structural Heart Intervention

OCT



TTE

CMR

CT



CCTA

Current  
Clinical  
Tasks

Coronary Vascular Anatomy  
Stent Dilatation

Anatomical Measurements  
3D Reconstruction

Diagnosis of Concomitant  
Coronary Artery Disease

AI  
Enhanced  
Pre-procedural  
Planning

Faster, Accurate & Reliable  
Higher-resolution

Non-invasive

# Prediction # 4 Emergence of Robotics + AI

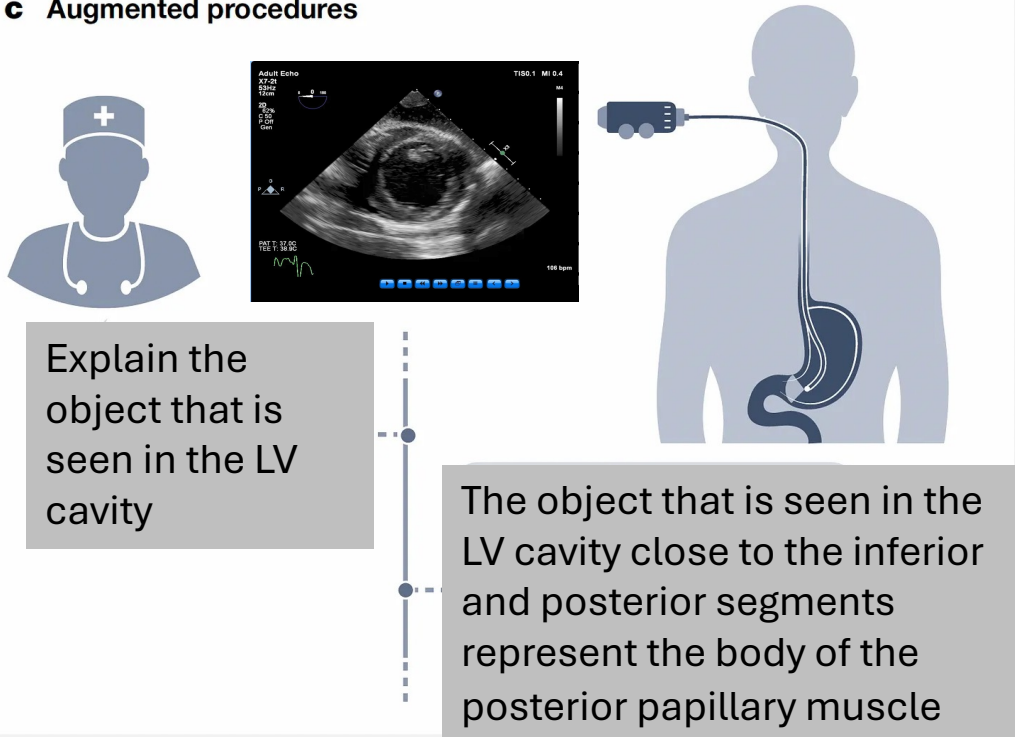
922

IEEE TRANSACTIONS ON MEDICAL ROBOTICS AND BIONICS, VOL. 4, NO. 4, NOVEMBER 2022

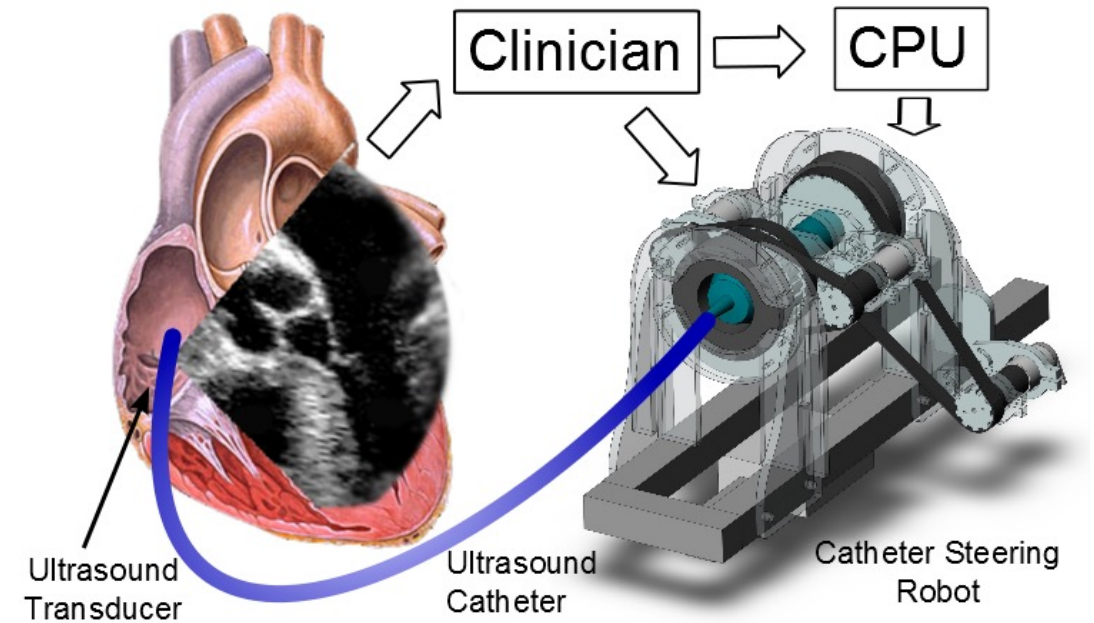
## Toward the Design and Development of a Robotic Transcatheter Delivery System for Mitral Valve Implant

Namrata U. Nayar<sup>1</sup>, Graduate Student Member, IEEE, Ronghuai Qi<sup>2</sup>, and Jaydev P. Desai<sup>3</sup>, Fellow, IEEE

### c Augmented procedures



*Nature* volume 616, pages259–265 (2023)



Demonstration of Individual Bending Modules of the Steerable Tip



# Prediction # 5 Newer Therapies and Interventions

## Prediction #4 New uses in Cardiovascular Therapies

Cardiac  
Surgery  
1953

Cardiac  
Transplant  
1967

Minimally Invasive  
Cardiac Surgery  
1990s

2025

3D-printed tissue  
Gene Therapies  
Cell therapies  
Nanoparticles  
Nanorobotics

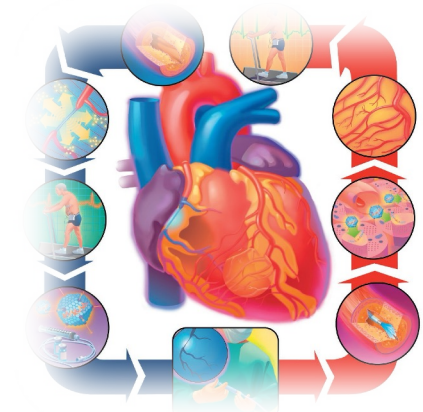
Cardiac  
Cath  
1927

Drugs  
CCU  
1962

Invasive  
Cardiology  
1977

Structural Heart  
Interventions  
2002

2030





# Software Agent

1977

1990

+

# GenAI

=

# Agentic AI

GAN

2014

Transformer

2017

GPT-2

2019

2020

Intelligent Agent

1995

Goal-directed AI

2010s

Autonomous AI

2014

AI Agent

2017

2024

[https://www.linkedin.com/posts/samerjoudi\\_artificialintelligence-agenticai-activity-7288646925984190464-ewiH](https://www.linkedin.com/posts/samerjoudi_artificialintelligence-agenticai-activity-7288646925984190464-ewiH)

nature

<https://doi.org/10.1038/s41586-025-09442-9>

Accelerated Article Preview

## The Virtual Lab of AI agents designs new SARS-CoV-2 nanobodies

On the morning of April 26, Tsinghua University held an inauguration ceremony for Tsinghua AI Agent Hospital and the 2025 Tsinghua Medicine Townhall Meeting at the Main Building Reception Hall. Tsinghua President Li Luming and Vice President Wang Hongwei attended the event.



# Digital Twin: Synthetic Generation of Echo from Surface ECG

nature cardiovascular research

Article

<https://doi.org/10.1038/s44161-025-00629->

## Synthetic generation of cardiac tissue motion from surface electrocardiograms

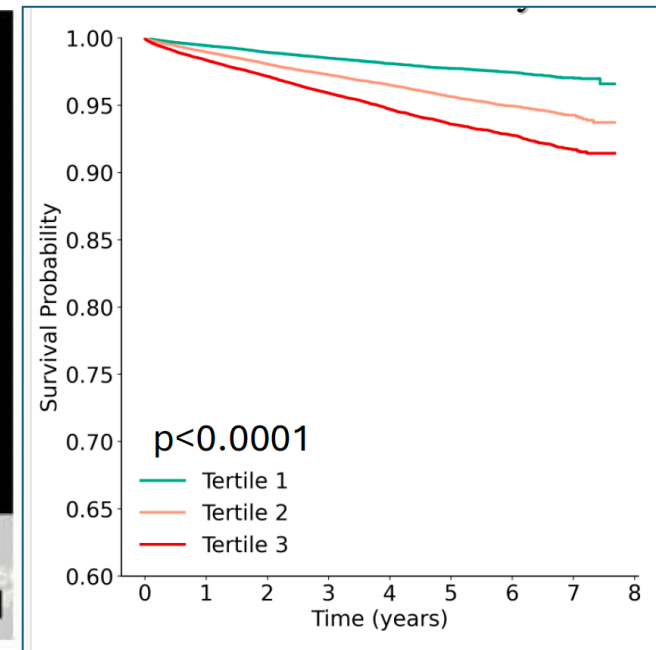
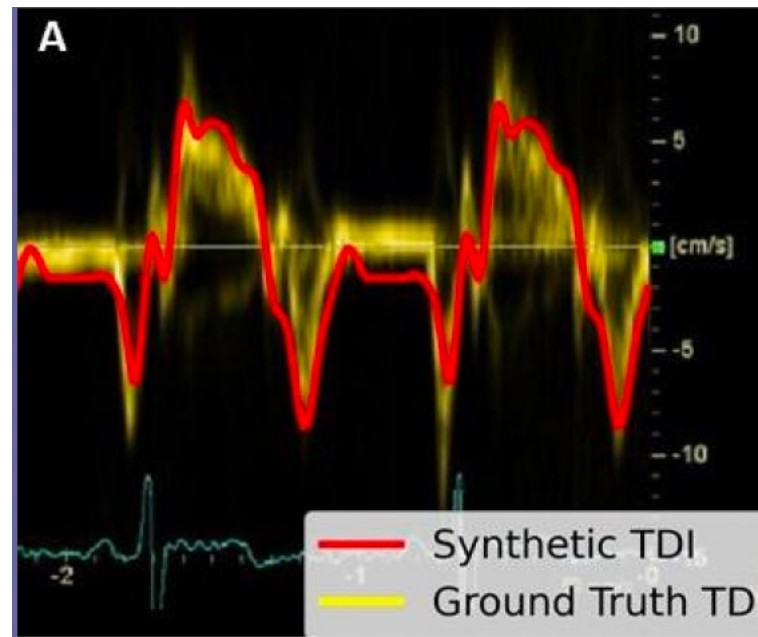
Received: 18 September 2024

Accepted: 27 February 2025

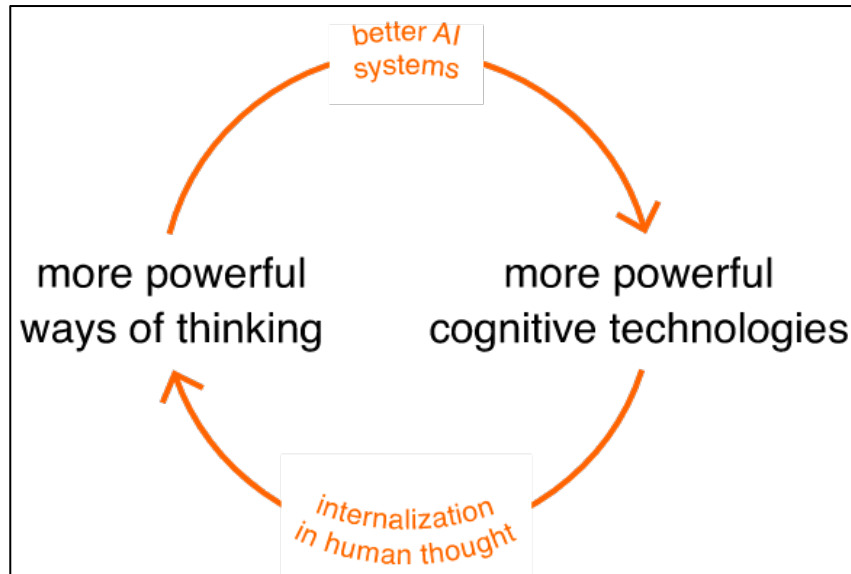
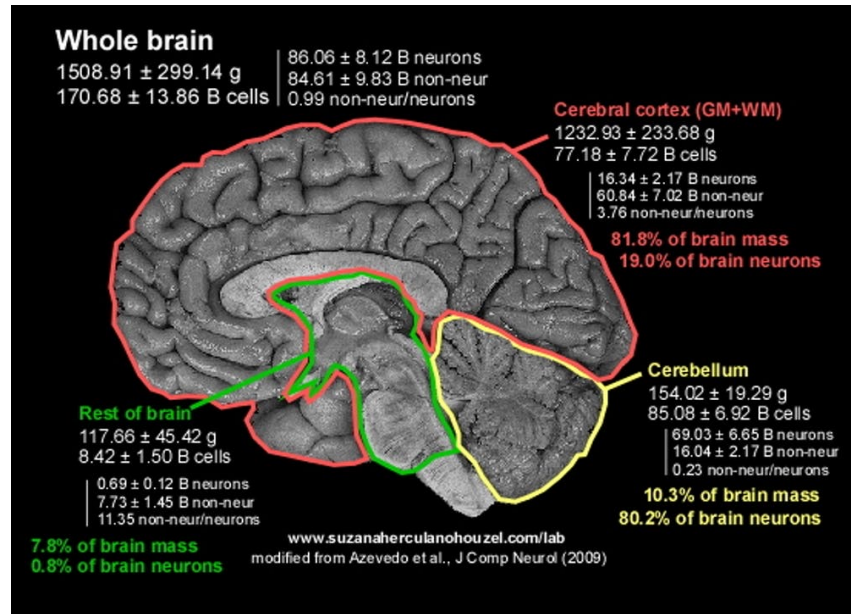
Published online: 14 April 2025

Aditya Radhakrishnan<sup>1,2,4</sup>, Naveena Yanamala<sup>2,3,4</sup>, Ankush Jamthikar<sup>3</sup>,  
Yanting Wang<sup>3</sup>, Sasha-Ann East<sup>3</sup>, Yasmin Hamirani<sup>3</sup>, Kameswari Maganti<sup>3</sup> &  
Partho P. Sengupta<sup>3</sup>✉

[Check for updates](#)

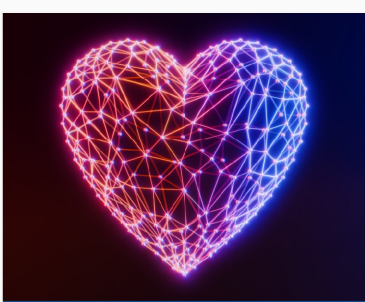


# AI in Cardiology



- We don't have the luxury of making 230 million years of mistakes and random variations to produce an intelligent machine **–AI should not replace critical thinking**
- “AI operates on a radically shorter time scale, with fewer degrees of freedom, takes a much more direct route to our objective.”  
**–Use AI for low-caliber activities**
- “AI bundles the objective facts from a data perspective” **–AI should not endanger empathy and morality in medicine**





Read more

Lancet Digital Health: Artificial intelligence and digital innovations in cardiovascular care

# Artificial Intelligence and Digital Innovations in Cardiovascular Care 1

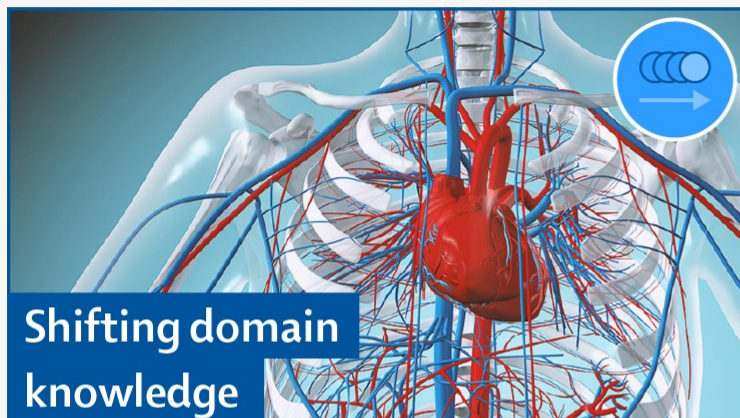


## Challenges for augmenting intelligence in cardiac imaging

Partho P Sengupta, Damini Dey, Rhodri H Davies, Nicolas Duchateau, Naveena Yanamala



### Barriers to the clinical implementation of AI technology



**Shifting domain knowledge**



**Human factors**



**Integration and implementation**



**Evaluation challenges**



**Scarcity of high-quality and standardised data**



**Clinical versus data science knowledge gap**



# New Generation of Leaders

## ASE Young Investigator Award Nominations & Winners



Josef Korinek  
2006



Mac Eleid  
2008



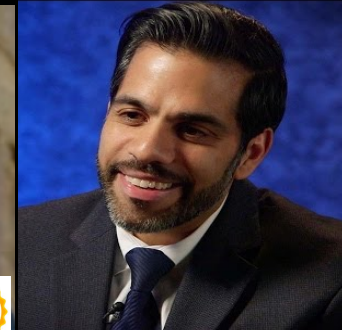
Nupoor Narula  
2010



Mayank Kansal  
2010



Sukrit Narula  
2015



Sanjeev Bhavnani  
2015



Ines Sheriffi  
2016



Marton Tokodi  
2018



Nobuyuki Kagiya  
2019



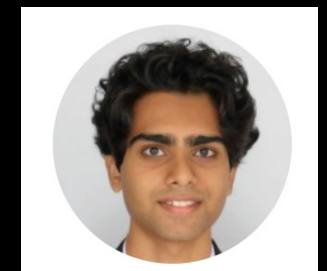
Ambarish Pandey  
2020



Quincy Hathaway  
2021



Rohan Shah  
2022



A Radhakrishnan  
2024

# Innovation & Strength-based Leadership

Center for innovation





Please evaluate this session  
using the QR Code

