

Advanced Imaging: What you Can't See, Will Hurt Your Patient

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DISCLOSURE

**Omid Jazaeri, MD, RPVI, FACS,
FSMB**

- No relevant financial relationship reported

Discovering the Inside

INNOVATIONS IN MEDICAL IMAGING AND RADIATION

This National Radiologic Technology Week, travel back in time and relive the history of the radiologic technology profession and the American Society of Radiologic Technologists through ASRT Museum and Archives exhibits.

1895

Commonly known as the first x-ray and within

1896

First clinical use of x-rays

1903

First use of radiography

1904

Thomas Edison's radiation exposure

1913

Known for his many contributions to science, William Coolidge invents the heated cathode, enabling external beam radiotherapy.

1914

Marie Curie designs vehicles to carry medical equipment to the battlefield in World War I.

1920

American Association of Radiologic Technicians established; known today as the American Society of Radiologic Technologists.

1927

Thirty-seven percent of all radiologic technologists are nurses.

1941

The U.S. military trains thousands of servicemen as radiographers to provide medical care for wounded soldiers overseas during World War II.

Madame Curie



“Good God, I have seen my death!”

Raymond V. Damadian



1977

Raymond V. Damadian, Lawrence Minkoff and Michael Teresi invent the first-ever MR image.

ASRT Museum and Archives

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2015

ASRT Museum and Archives opens.

2011

First clinical 3-D mammography exam performed.

1992

Congress enacts the Mammography Quality Standards Act ensuring all women have access to quality mammography.

1983

ASRT opens new national office in Albuquerque, New Mexico.

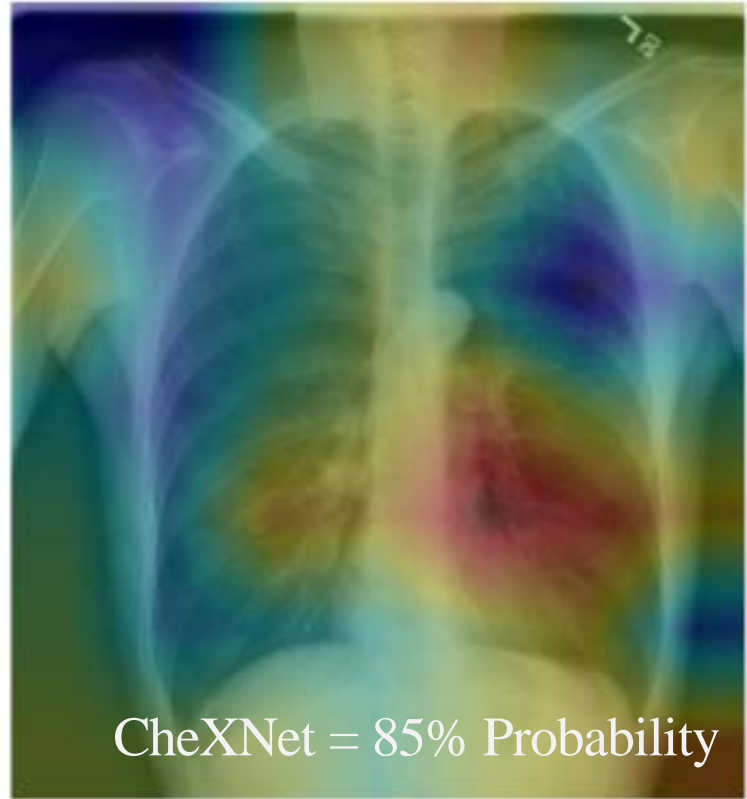


1979

Allan Cormack and Godfrey Hounsfield awarded Nobel Prize in Medicine for invention of computed tomography.



Artificial Intelligence in Medicine & Imaging



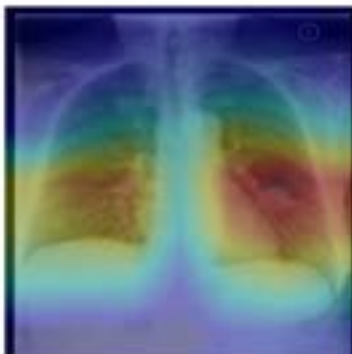
CheXNet = 85% Probability

Che

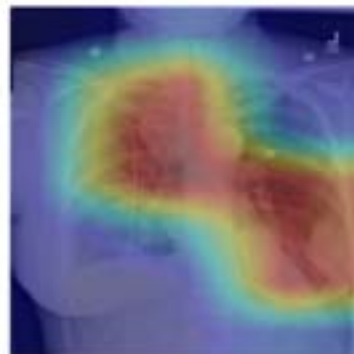
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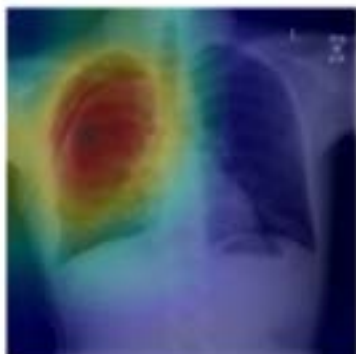
(a) Patient with multifocal community acquired pneumonia. The model correctly detects the airspace disease in the left lower and right upper lobes to arrive at the pneumonia diagnosis.



(b) Patient with a left lung nodule. The model identifies the left lower lobe lung nodule and correctly classifies the pathology.



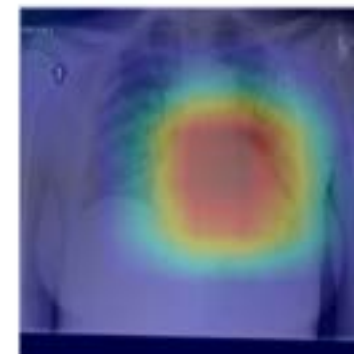
(c) Patient with primary lung malignancy and two large masses, one in the left lower lobe and one in the right upper lobe adjacent to the mediastinum. The model correctly identifies both masses in the X-ray.



(d) Patient with a right-sided pneumothorax and chest tube. The model detects the abnormal lung to correctly predict the presence of pneumothorax (collapsed lung).



(e) Patient with a large right pleural effusion (fluid in the pleural space). The model correctly labels the effusion and focuses on the right lower chest.



(f) Patient with congestive heart failure and cardiomegaly (enlarged heart). The model correctly identifies the enlarged cardiac silhouette.

Conclusion

- *“We developed an algorithm which **exceeds the performance of radiologists in detecting pneumonia from frontal-view chest X-ray images**. We also show that a simple extension of our algorithm to detect multiple diseases outperforms previous state of the art on ChestX-ray14, the largest publicly available chest X-ray dataset....”*

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Artificial Intelligence: What you Can't Interpret, Will Hurt Your Patient

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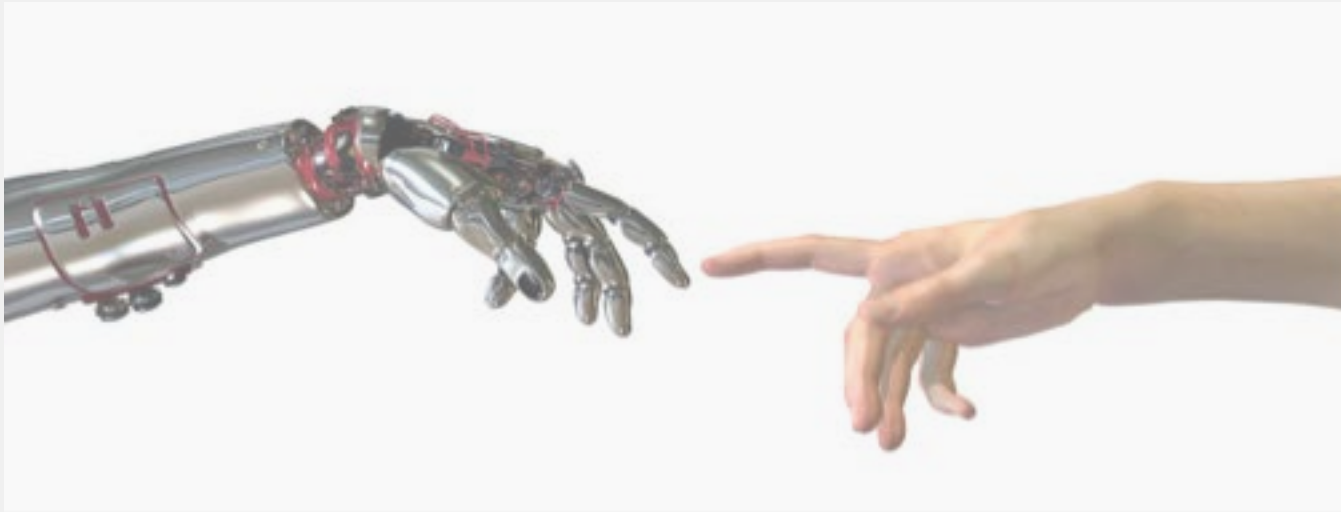
Adjunct Clinical Faculty, Department of Bioengineering



University of Colorado **Anschutz Medical Campus**

Artificial Intelligence

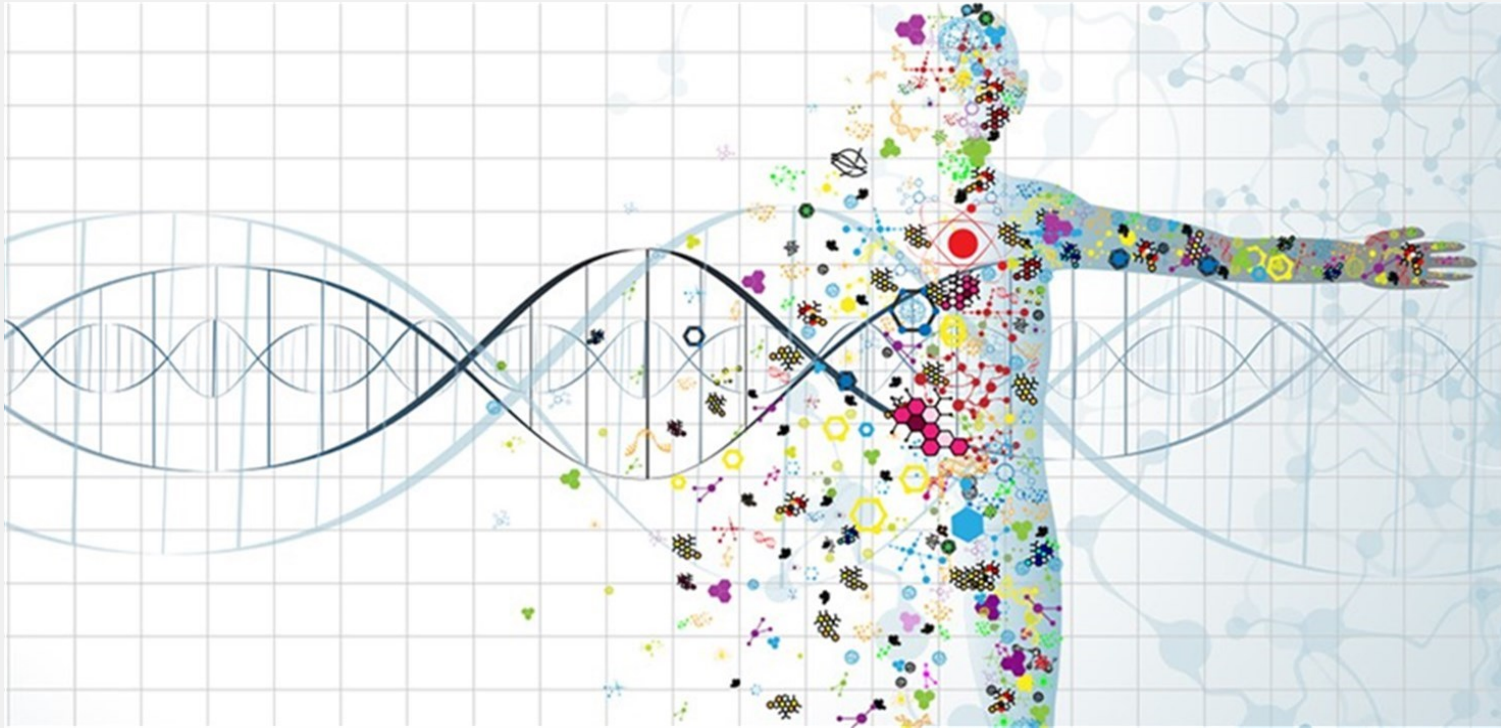
- **Definition:** Any device that perceives its environment and takes actions that maximize its chance of success at some goal.



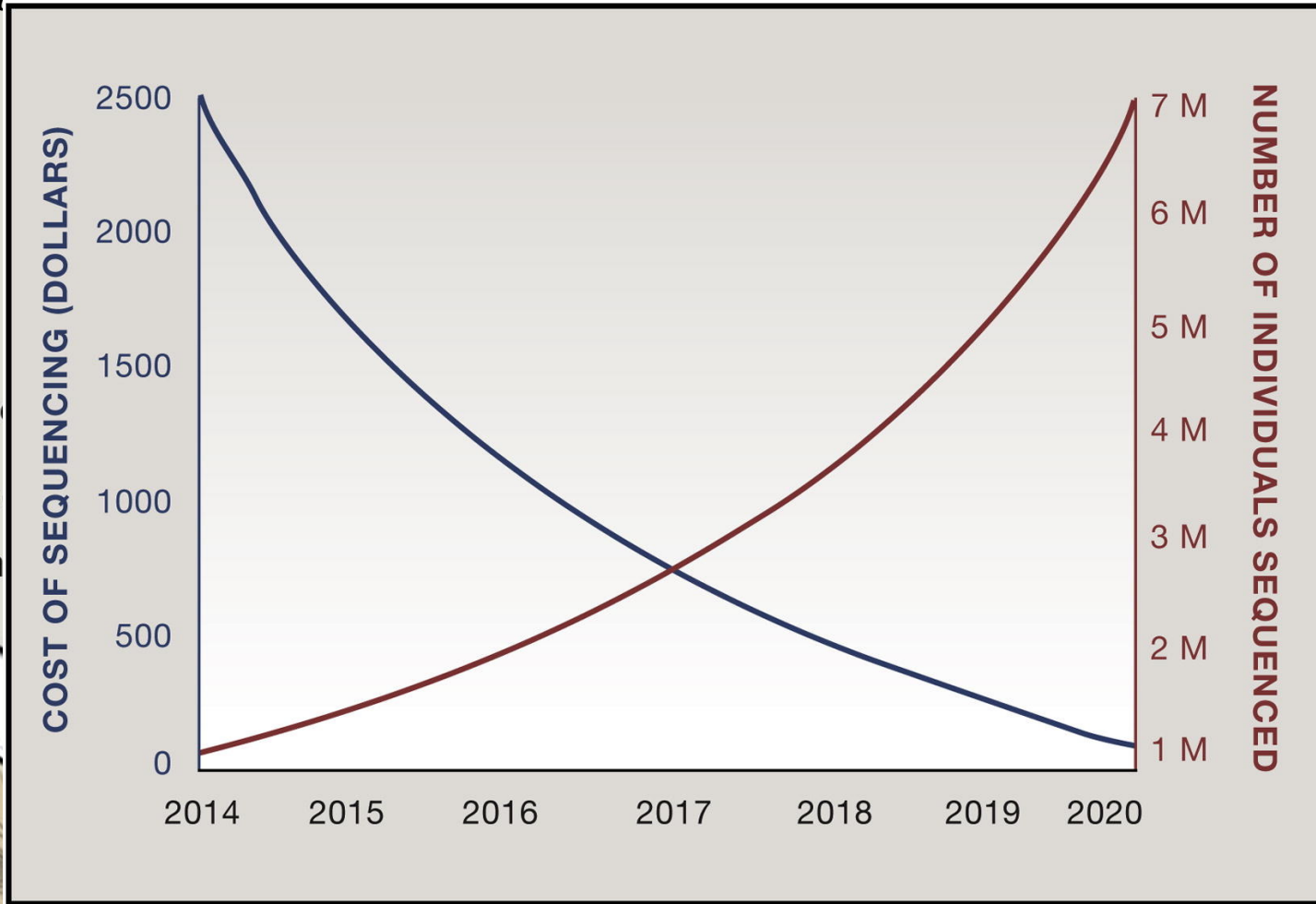
Artificial Intelligence



Artificial Intelligence, Big-Data, Machine learning and “Deep Learning”



Individualized Medicine from Pre-womb to Tomb



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Epigen
Exposom

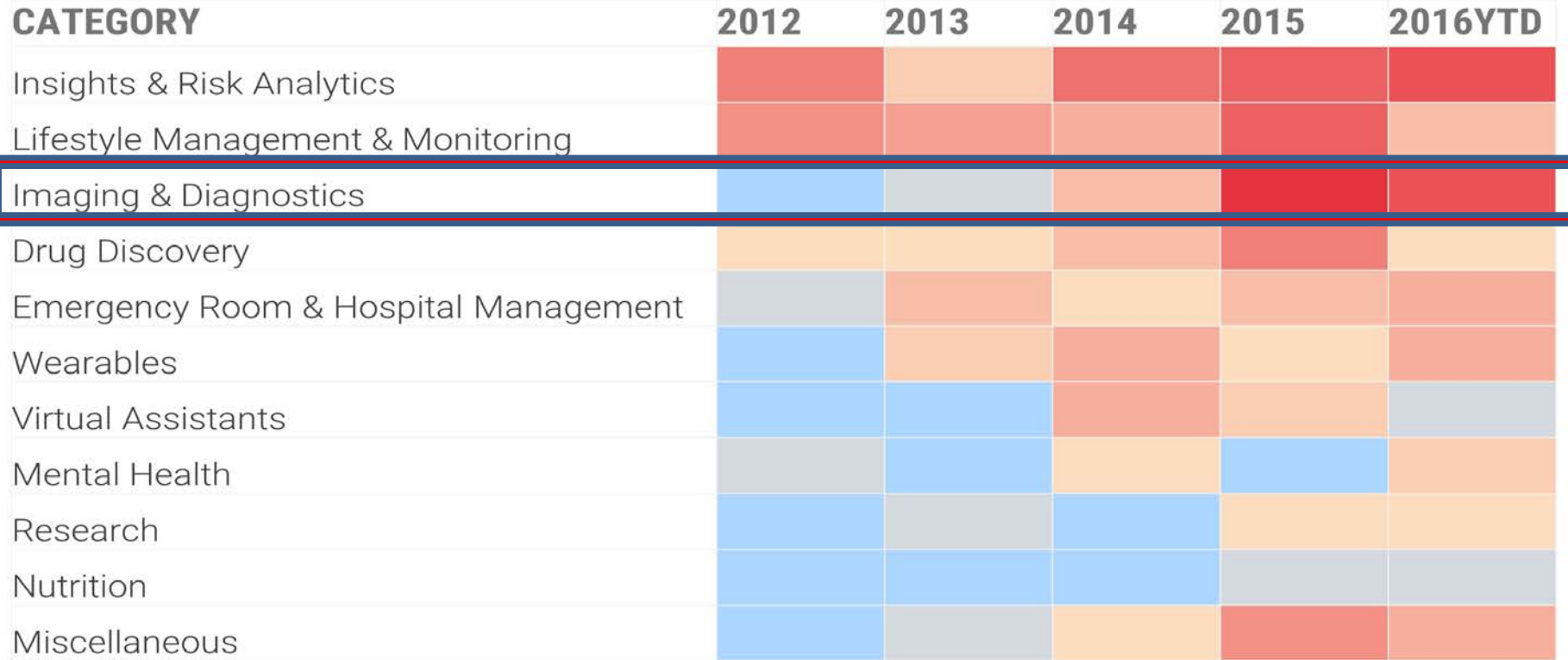


l graph
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AI IN HEALTHCARE: CATEGORY HEATMAP

2012-2016 (as of 8/30/2016)



<https://www.cbinsights.com/research/artificial-intelligence-healthcare-investment-heatmap/>

The Problem With Big-Data

- Algorithms used biased data
 - » Racial, religious, gender-general
 - » Education, location, profession-specific
- Transparency
 - » Show your work
- Accountability
 - » AI vs Human

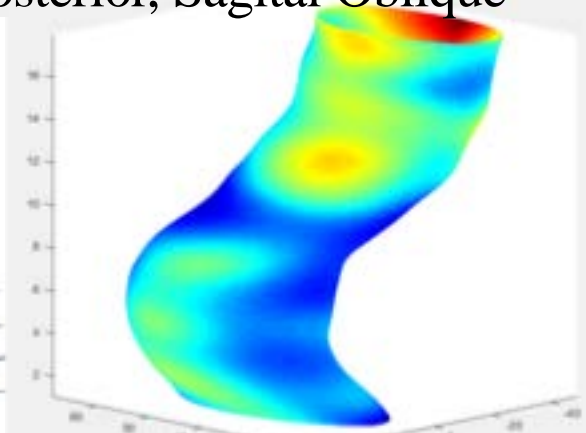
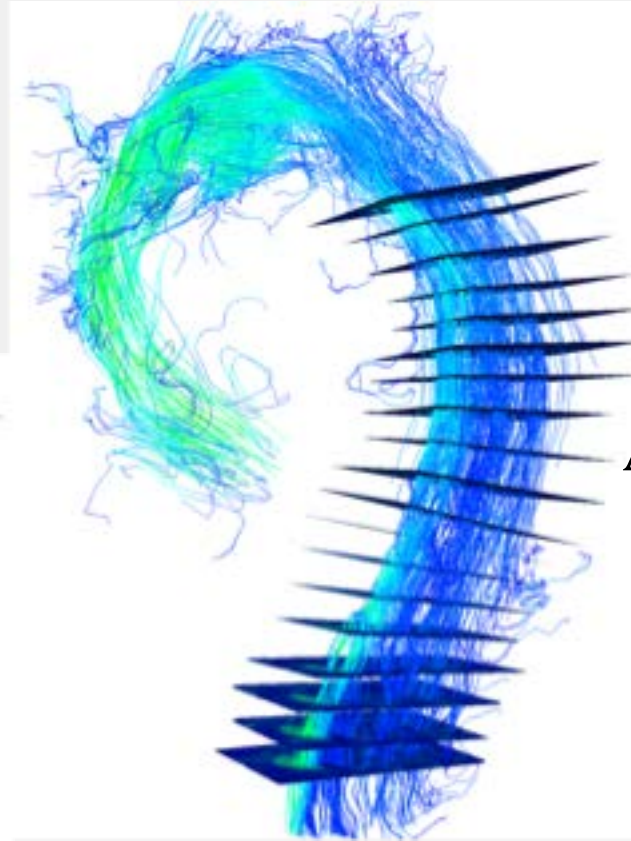
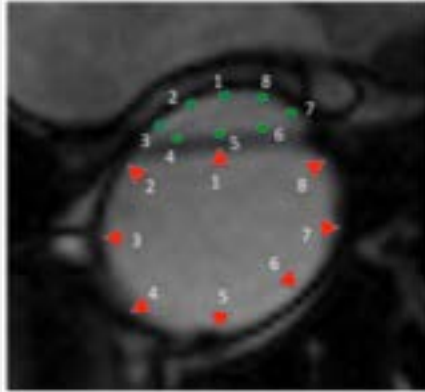


“Deep Learning”

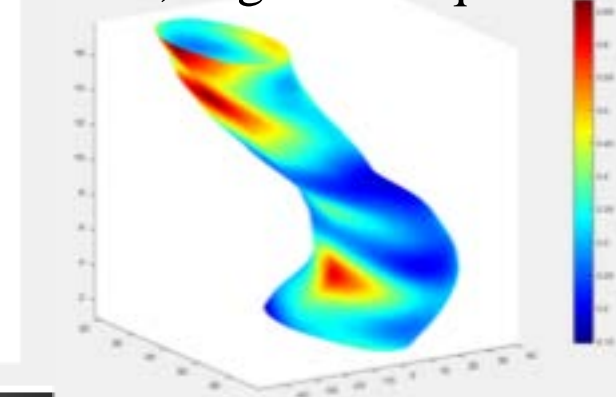
- Medical Applications in Imaging
 - » Diagnostic Radiology
 - » Tumor Detection-Pathology
 - » Tracking Tumor Development-Pathology
 - » Diabetic Retinopathy
 - » Genomics-Personalized Medicine
 - » **Blood Flow Quantification and Blood Vessel Visualization**

Blood Flow Quantification and Visualization

Posterior, Sagittal Oblique



Anterior, Sagittal Oblique



In-plane WSS	True Lumen	False Lumen
a	0.27 (0.16 - 0.38)	0.07 (0.05 - 0.11)
b	0.33 (0.27 - 0.36)	0.10 (0.07 - 0.13)
c	0.33 (0.25 - 0.39)	0.10 (0.05 - 0.14)
d	0.31 (0.24 - 0.41)	0.14 (0.11 - 0.16)
e	0.29 (0.24 - 0.43)	0.09 (0.07 - 0.16)
f	0.26 (0.21 - 0.32)	0.10 (0.07 - 0.11)
g	0.25 (0.22 - 0.28)	0.09 (0.07 - 0.11)
h	0.30 (0.23 - 0.38)	0.11 (0.09 - 0.13)
i	0.24 (0.18 - 0.34)	0.08 (0.07 - 0.12)
j	0.19 (0.15 - 0.25)	0.07 (0.05 - 0.11)
k	0.25 (0.17 - 0.34)	0.09 (0.07 - 0.12)
l	0.34 (0.28 - 0.43)	0.11 (0.08 - 0.15)
m	0.37 (0.31 - 0.54)	0.10 (0.08 - 0.14)
n	0.44 (0.33 - 0.50)	0.11 (0.10 - 0.12)
o	0.39 (0.32 - 0.58)	0.09 (0.06 - 0.14)
p	0.32 (0.27 - 0.39)	0.12 (0.09 - 0.13)
q	0.37 (0.28 - 0.55)	0.09 (0.06 - 0.12)
r	0.44 (0.38 - 0.50)	0.11 (0.09 - 0.14)

Association of variant arch anatomy with type B aortic dissection and hemodynamic mechanisms

Sherene Shalhoub MD, MPH, FACS^a, Michel Schäfer MSc^b, Thomas S. Hatsukami MD, FACS,^a

Matthew

Susanna

Benjamin

and Dubin

RCSI,^c

Colorado, Colo;

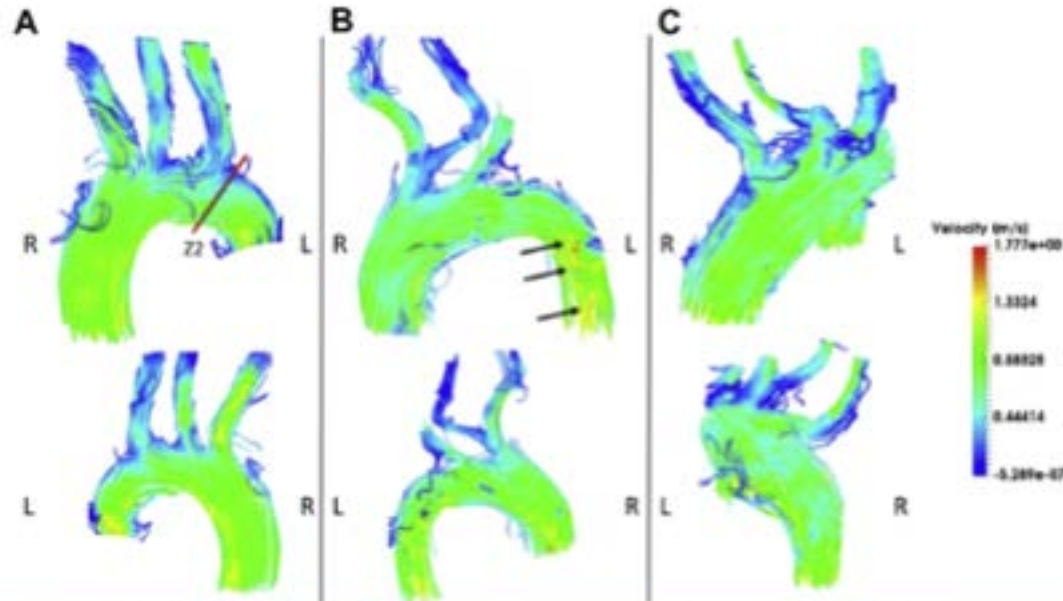


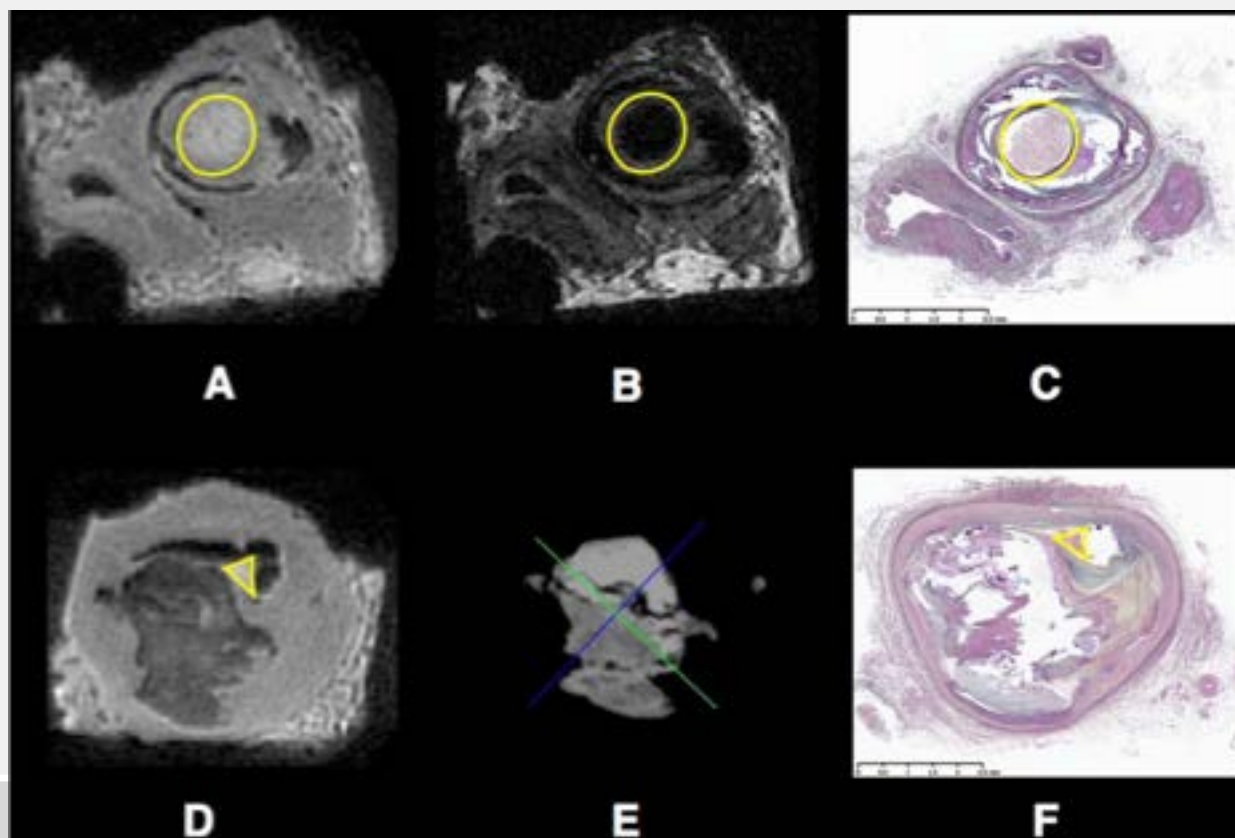
Fig 2. Streamline visualization of three different aortic arch morphologies from three healthy controls via four-dimensional flow magnetic resonance imaging (4D flow MRI). **(A)** Conventional aortic arch revealing uniform velocity distribution at the distal portion of the aortic arch. **(B)** Bovine aortic arch presented significant flow acceleration at the region of the inner curve (arrows), creating elevated regional wall shear stress. **(C)** Aberrant right subclavian aortic arch revealed relatively uniform velocity distribution.

2018 May 24.

Peripheral Endovascular Interventions in the Era of Precision Medicine: Tying Wire, Drug, and Device Selection to Plaque Morphology

Journal of Endovascular Therapy
2016, Vol. 23(5) 751-761
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DOI: 10.1177/1526602816653221
www.jevt.org
SAGE

Trisha Roy, MD^{1,2}, Andrew D. Dueck, MD, MSc^{1,2}, and Graham A. Wright, PhD^{1,3}



Thin Cap Fibroatheroma

- Covered stent
- Excimer laser
- Rotational/aspiration atherectomy



Fibrous Plaque

- Cutting/scoring balloon



Pathologic Intimal Thickening

- Drug eluting technology
- Cryoplasty



Chronic Total Occlusion

- Excimer laser
- Orbital atherectomy
- Directional atherectomy
- Collagenase



Diffuse Speckled Calcium

- Directional atherectomy
- Orbital atherectomy
- Collagenase



Acute Thrombus/Calcified Nodule

- Excimer laser
- Rotational/aspiration atherectomy



Concentric Calcium

- Orbital atherectomy
- Directional atherectomy



collagen



healed thrombus/
dense fibrous tissue



calcium



necrotic core

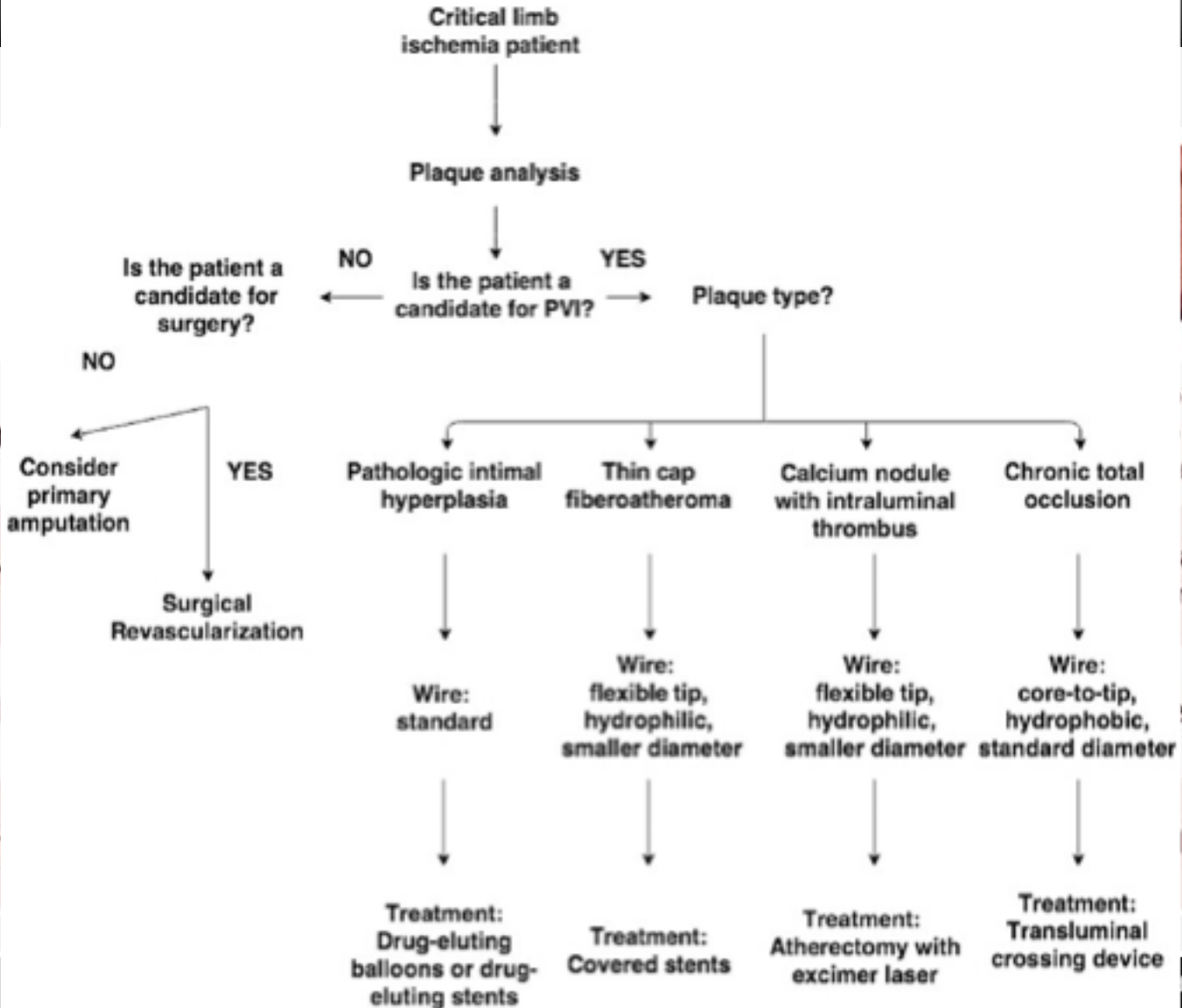


smooth muscle

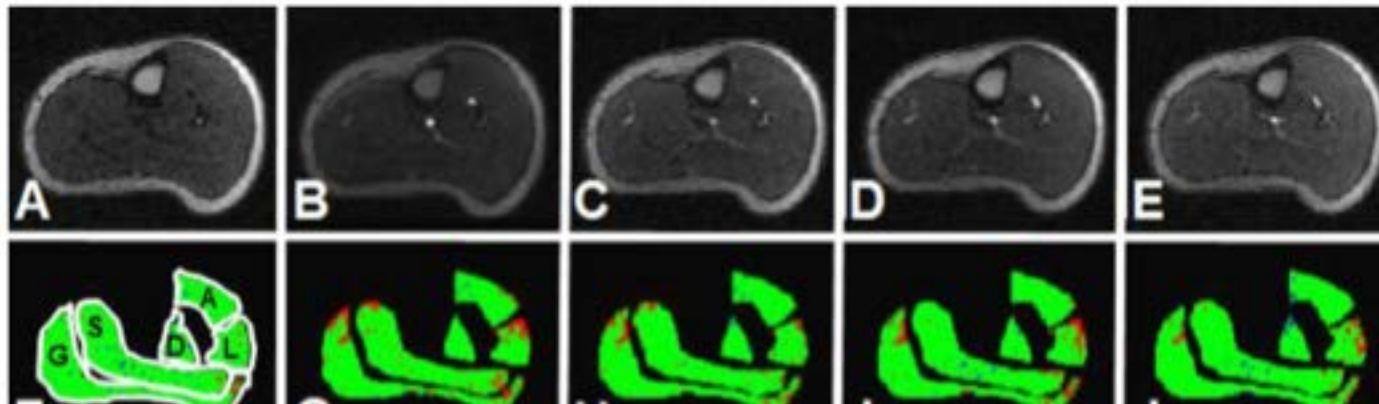


acute thrombus

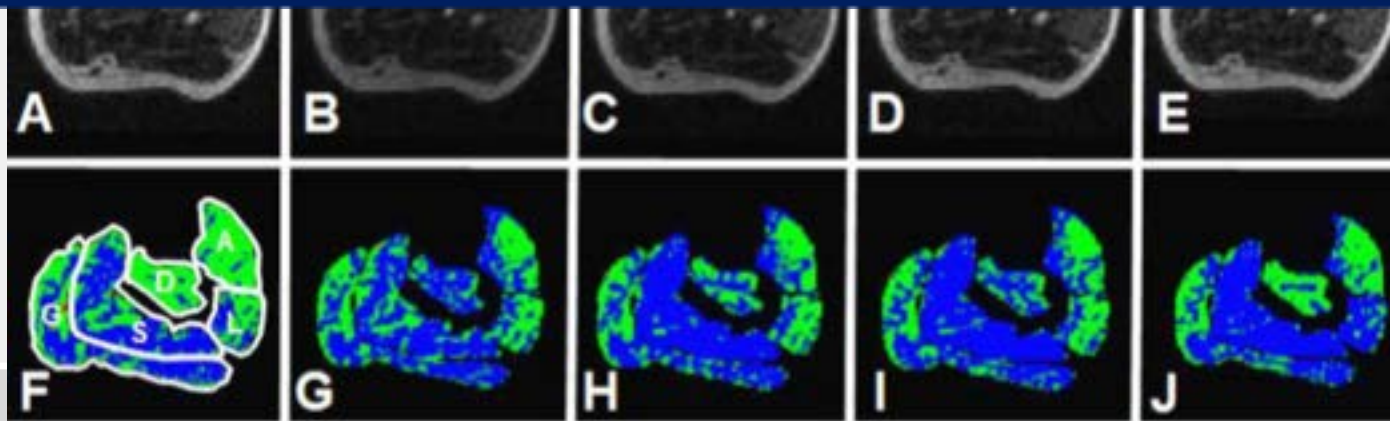
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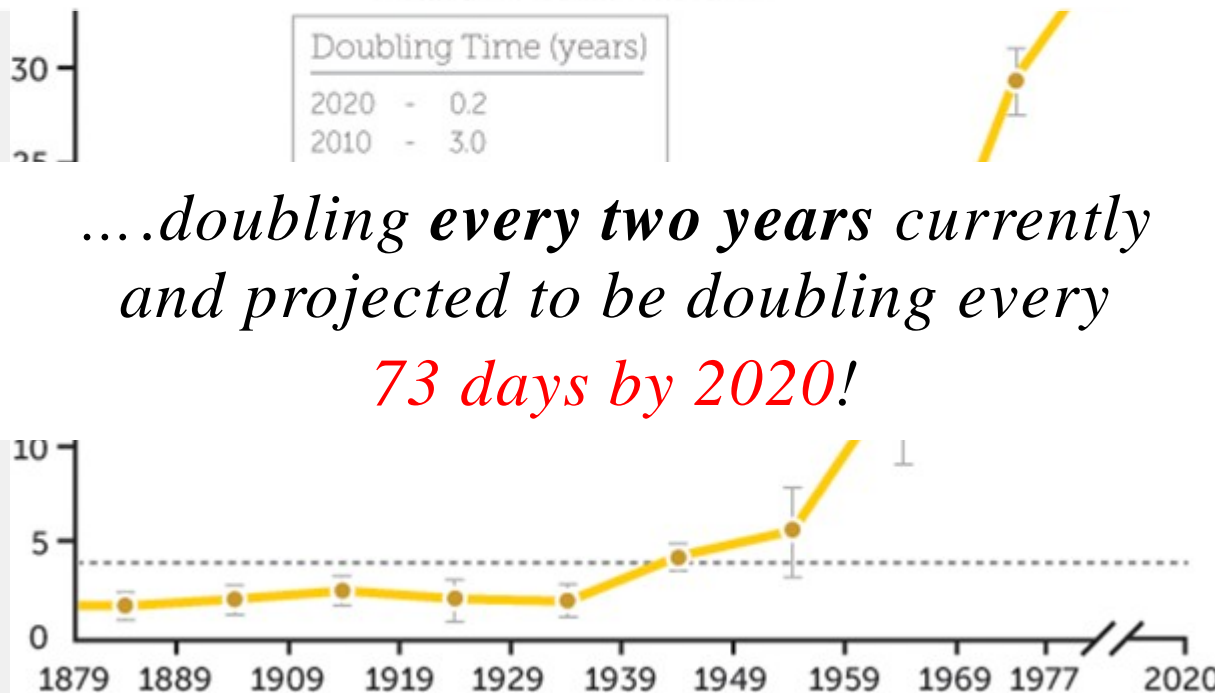
For PAD patients: hypointense voxel fractions of AM, LM, GM were inversely correlated with eGFR ($r = -0.509$, $p = 0.008$; $r = -0.441$, $p = 0.024$; $r = -0.431$, $p = 0.028$).



CHALLENGES AND OPPORTUNITIES FACING MEDICAL EDUCATION

PETER DENSEN, MD

IOWA CITY, IOWA



Transactions of The American Clinical and Climatological Association in 2011; 122: 48-58

How much effort is needed to keep up with the literature relevant for primary care?

*By Brian S. Alper, MD, MSPH
editor@dynamicmedical.com
Research Assistant Professor*

- 341 Journals
- 7287 articles published monthly
- 627.5 hours/month
- 87% of total available time in a month spent reviewing literature
- Only 3hours/day are left.....

JAMA | **Original Investigation**

Diagnostic Assessment of Deep Learning Algorithms for Detection of Lymph Node Metastases in Women With Breast Cancer

Babak Ehteshami Bejnordi, MS; Mitko Veta, PhD; Paul Johannes van Diest, MD, PhD; Bram van Ginneken, PhD; Nico Karssemeijer, PhD; Geert Litjens, PhD; Jeroen A. W. M. van der Laak, PhD; and the CAMELYON16 Consortium

Conclusion

- » In cross-sectional analyses that evaluated 32 algorithms submitted as part of a challenge competition, 7 deep learning algorithms showed greater discrimination than a panel of 11 Pathologists with an AUC of 0.994 (best algorithm) vs. 0.884 (best pathologist)

Summary

- Our current imaging paradigms are outdated at best
- AI is here to stay
- Deep Learning is important but task focused
- Avoid “trial and error”
- Know your own “Deep Blue” moment
- Take a break from drugs and devices



Thank You!

