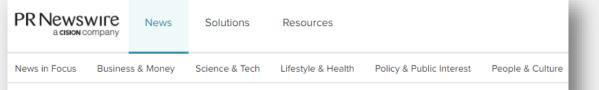


Machine Learning in Radiology

Amir Tahmasebi, PhD Senior Scientist Philips Research North America



Arterys Receives 510(k) Clearance for Arterys Software for Cloud-Based Medical Image Visualization and Quantification

First Comprehensive Visualization and Quantification of Cardiac 4D Flow and cardiac Function Available to Physicians in clinical settings



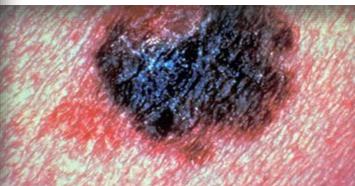
This Al can spot skin cancer as accurately as a doctor

Cancer

WIRED

Artificial Intelligence

The artificial intelligence was trained on an image database of 129,000 images and performed as well as trained medical professionals



Radiology by robots: this is what breast cancer looks like to tumour-hunting Al

Science

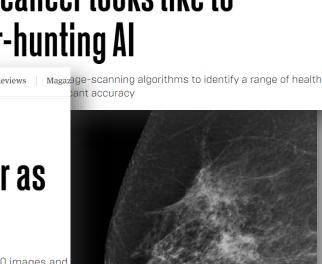
Culture

Video

Reviews

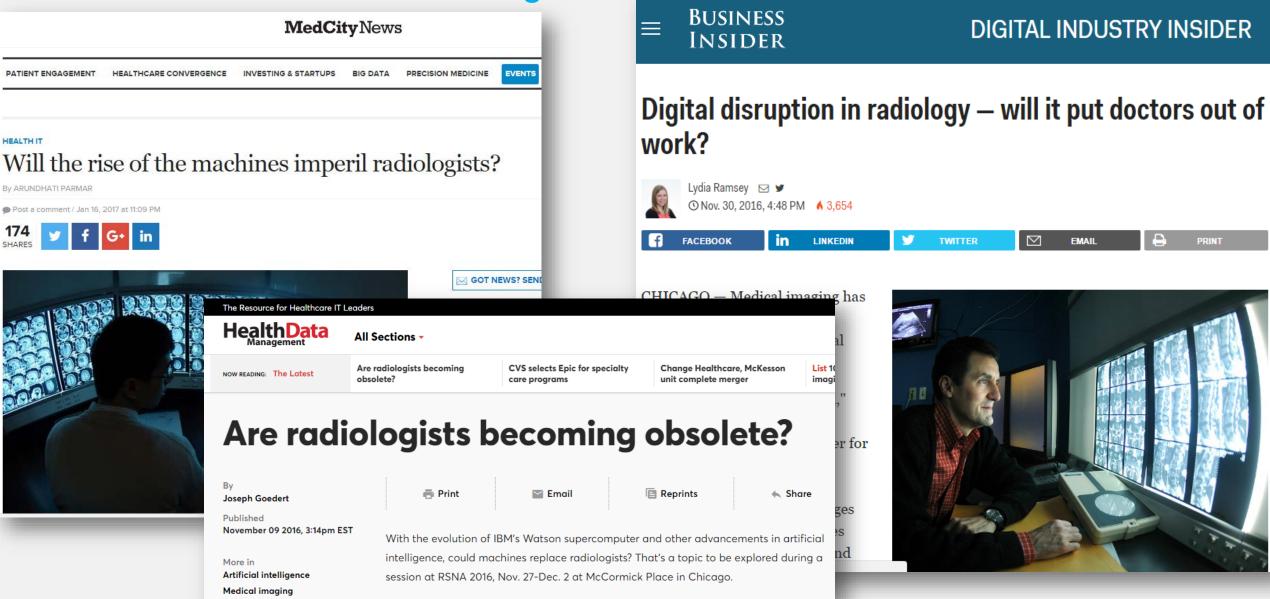
Technology Video Reviews Science Culture ant accuracy

Technology



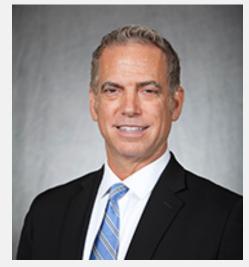
Magazine

human versus machine learning



human + machine learning





Keith Dreyer, DO, PhD, vicechairman, radiology, computing and information sciences, Massachusetts General Hospital

Securing Radiology's Future Through the Digital Revolution

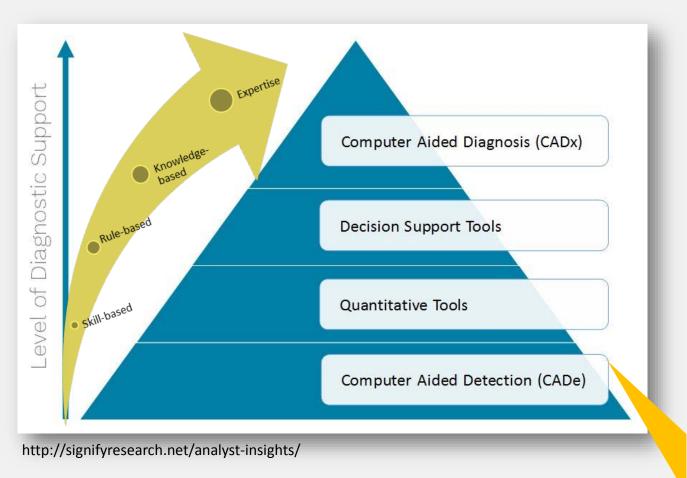
The most effective use for AI is one using a Centaur approach where computers and humans work together, Dreyer said.

News | November 28, 2016 By Julie Kaufield, MA, RT(R)

"To make new radiology, you have to understand our domain, have access to our environments, and have a vision of what radiology is going to look like tomorrow," Dreyer said, regarding the value radiologists bring to machine learning.

Radiology of the future will be about using machines to make us smarter, do more, and give us more value by giving us more time to communicate, he said.

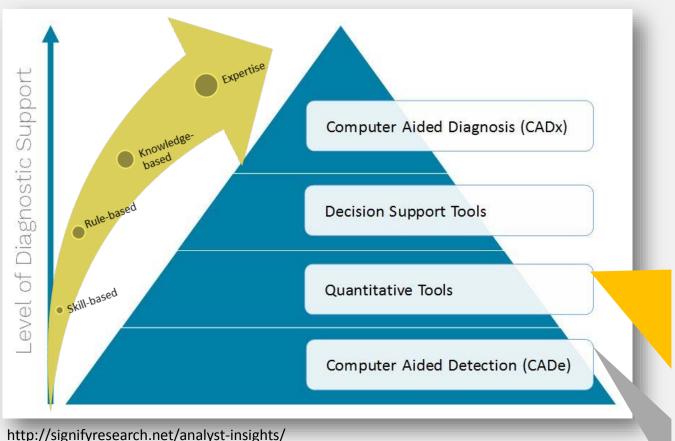
How/Where can it be useful?



- Automation
- Accuracy
- Consistency

Identify a variety of cancers such as breast cancer, prostate cancer, and lung lesions

How/Where can it be useful?



• Automation

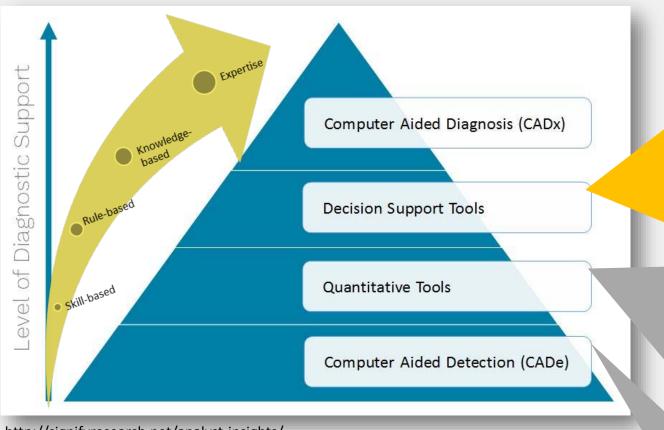
- Accuracy
- Consistency

automatic detection and measurements of imaging features (biomarkers) to assist with diagnosis, such as lung density, breast density, analysis of coronary and peripheral vessels, etc.

4D Flow from Arterys

Identify a variety of cancers such as breast cancer, prostate cancer, and lung lesions

Artificial Intelligence for Radiology *How/Where can it be useful?*



http://signifyresearch.net/analyst-insights/

- Integration
- X-collaboration

detection and quantification, alongside supporting information extracted from an EHR, pathology reports and other patient records, to assist with diagnosis IBM Watson Health

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How/Where can it be useful?

Multi-modal Data

- Image
 - Pixel/voxel
- Non-image
 - EMR
 - Genomics
 - ...



Artificial Intelligence for Radiology *How/Where can it be useful?*

Collaboration

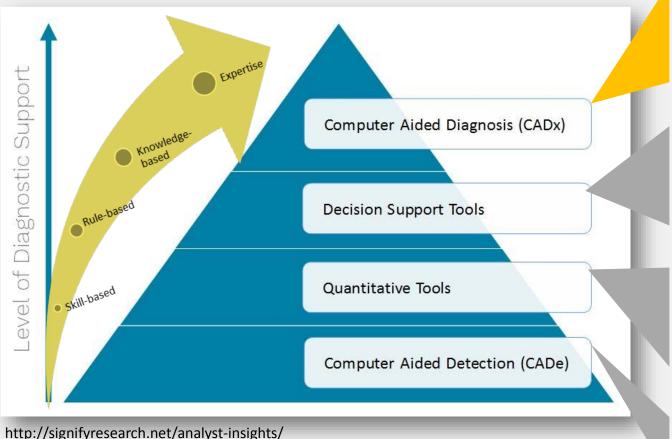
...

- One radiologist vs. multiple radiologists
- Radiologist + Pathologist
- Radiologist + Oncologist





How/Where can it be useful?



provide information beyond detection and quantification by also providing interpretation of the scan, for example by providing a probability score for the presence of cancer Aidence, Enlitic and CureMetrix

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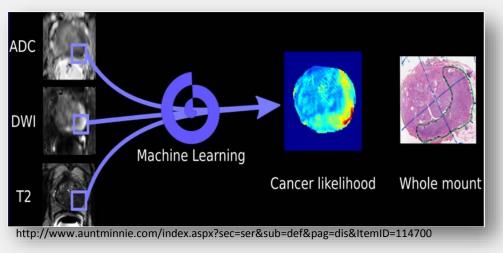
4D Flow from Arterys

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iCAD

How/Where can it be useful?

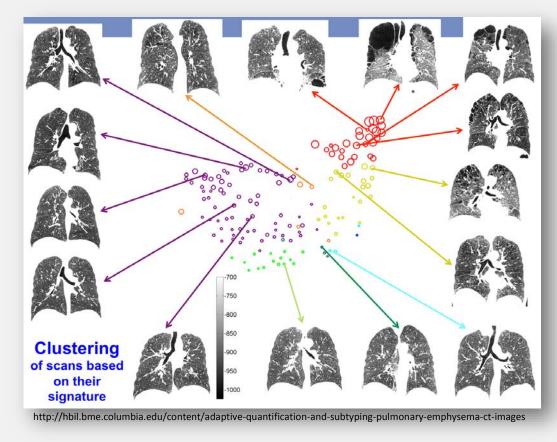
- Supervised
 - Learning from [human] expert



- PIRADS
- BIRADS
- LUNGRADS
- •

...

- Unsupervised
 - Discover new knowledge



How/Where can it be useful?

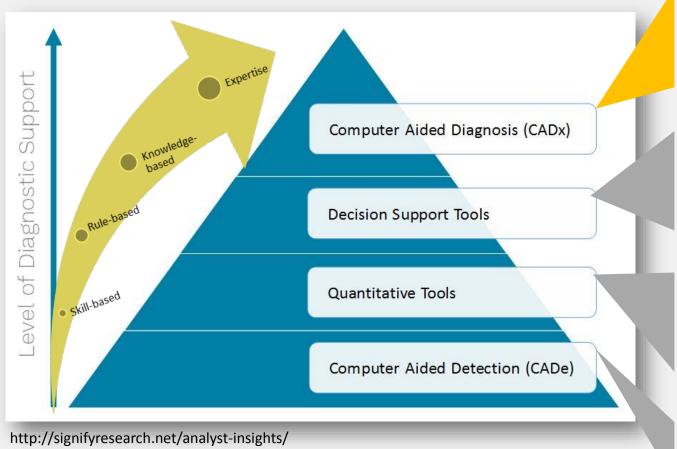
- Human
- n VS Machine VS (Human + Machine)



Human + Machine > Human/Machine

https://www.datanami.com/2015/10/16/avoiding-the-pitfalls-of-bigger-data-at-the-human-machine-interface/

How/Where can it be useful?



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4D Flow from Arterys

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iCAD

Artificial Intelligence for Radiology *Challenges*

- Supervised
 - [Annotated] data as Ground Truth
 - Domain dependent
 - Scalability

Unsupervised

- Lucky-guess
- Scalability

Artificial Intelligence for Radiology The danger

- Can a machine think by itself and come up with new rules?
 - intuit unexpected insights,
 - conjure alternative scenarios
 - understand emotion

Trained machines are exquisitely well suited to their environment—and ill-adapted to any other.

- University of Pittsburgh Medical Center
- **Goal:** using machine learning to predict whether pneumonia patients might develop severe complications
 - to send patients at low risk for complications to outpatient treatment, preserving hospital beds and the attention of medical staff
 - The model did what it was told to do: Discover a true pattern in the data.
 - One of the rules instructed doctors to send home pneumonia patients who already had asthma, despite the fact that asthma sufferers are known to be extremely vulnerable to complications.
 - hospital policy to send asthma sufferers with pneumonia to intensive care

Artificial Intelligence for Radiology *Role of Industry*

• Facilitate integration to workflow

Right information at the right time

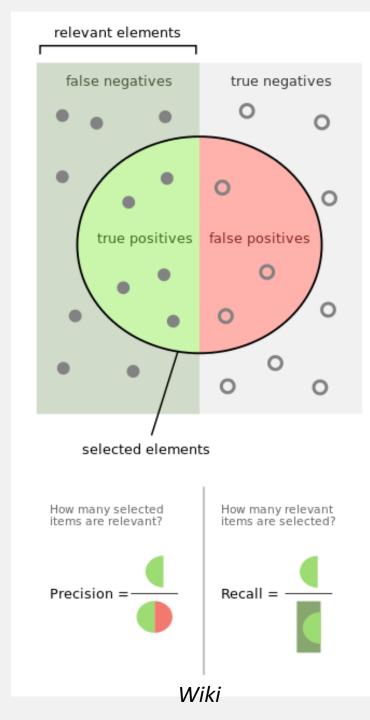


What matters from clinician's point of view

- Precision / Recall trade-off
 - Predict cancer with high confidence
 - Avoid missing too many cancer cases

$$Precision = \frac{TP}{TP + FP}$$
$$Recall = \frac{TP}{TP + FN}$$

 $F = 2 \times \frac{Precision \times Recall}{Precision + Recall}$

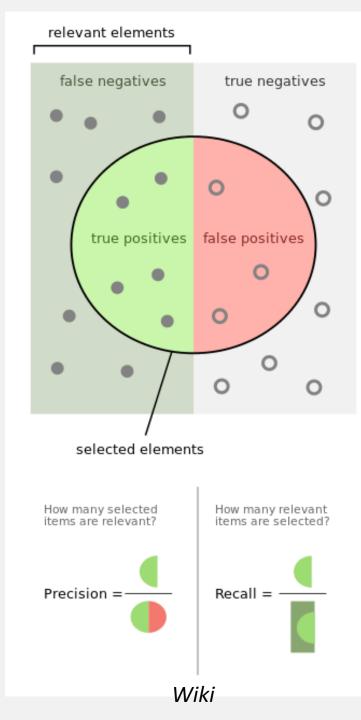


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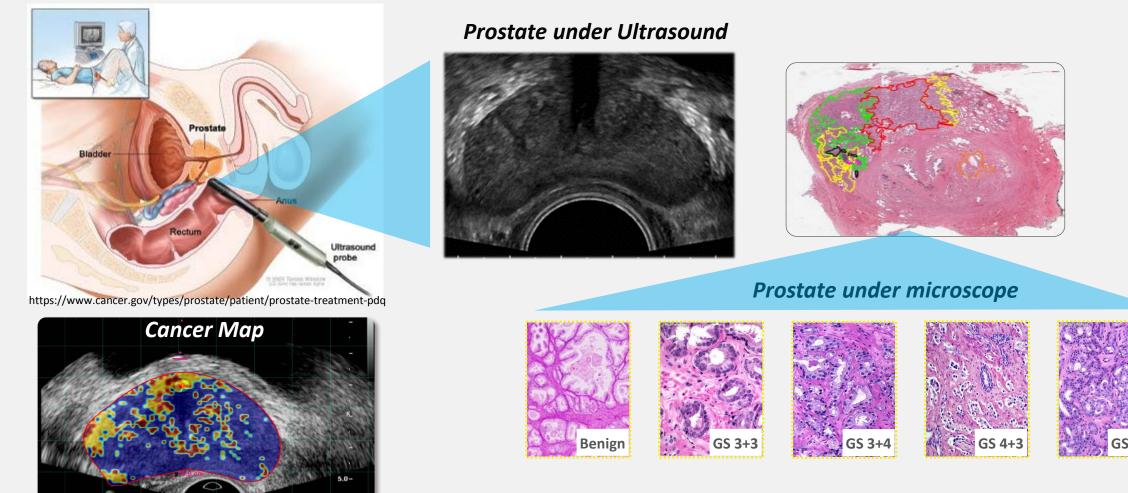
$$Specificity = \frac{TN}{TN + FP}$$
$$Sensitivty = \frac{TP}{TP + FN}$$



 $F = 2 \times \frac{Precision \times Recall}{Precision + Recall}$

Use Case: Tumor Tissue Characterization using Ultrasound

Problem Definition:



Courtesy of Dr. Abolmaesumi

Use Case: Tumor Tissue Characterization using Ultrasound

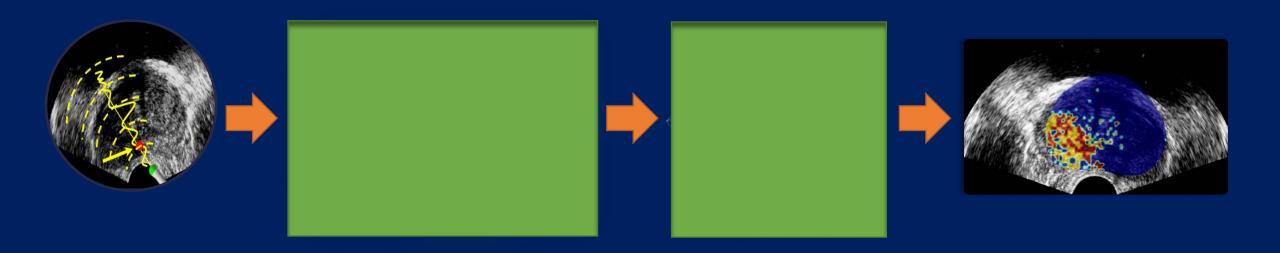
Steps:

- 1. Understand the problem
- 2. Define input(s) and output(s)
- 3. Investigate limitations and boundary conditions
- 4. Collect representative data
 - 1. [Labels]
- 5. [Calculate features/engineer features]
- 6. Define ML framework
- 7. Define Metric for evaluation

Use Case: Tumor Tissue Characterization using Ultrasound

Steps:

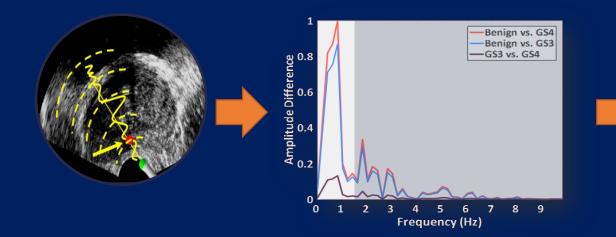
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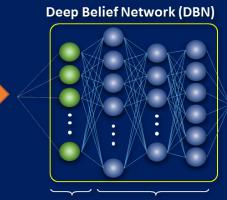


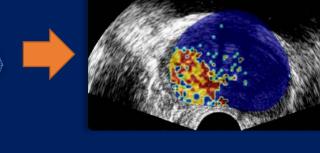
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Visible Layer Hidden Layers

Courtesy of Dr. Abolmaesumi

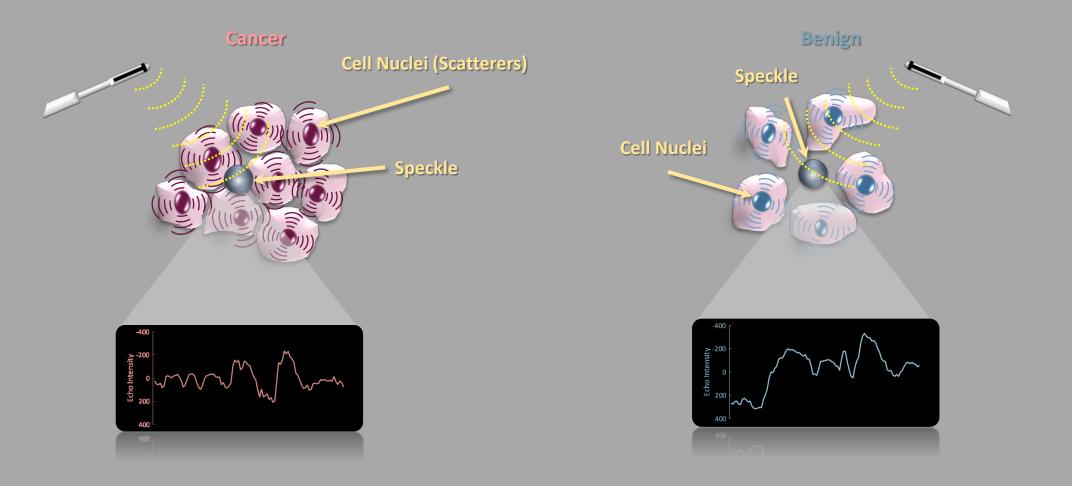
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Interpretability

Use Case: Tumor Tissue Characterization using Ultrasound



Courtesy of Dr. Abolmaesumi

References

[1] Classifying cancer grades using temporal ultrasound for transrectal prostate biopsy, International Conference on Medical Image Computing and Computer-Assisted Intervention (MICCAI), 653-661, 2016

[2] Ultrasound-based detection of prostate cancer using automatic feature selection with deep belief networks, International Conference on Medical Image Computing and Computer-Assisted Intervention (MICCAI), 70-77, 2015

[3] Ultrasound-based predication of prostate cancer in MRI-guided biopsy, Workshop on Clinical Image-Based Procedures, 142-150, 2014

Questions and Comments?

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