

Clinical Artificial Intelligence

Dr Nicholas Fuggle

Associate Professor in Rheumatology

Honorary Consultant Rheumatologist

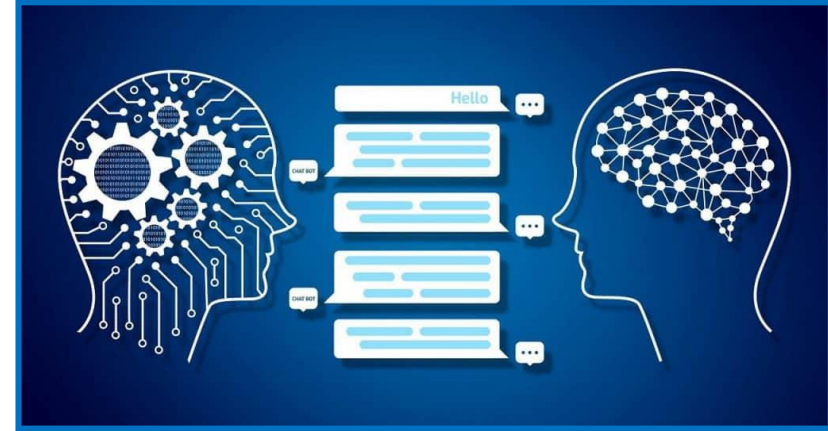
Clinical AI Interest Group Co-Organiser

Clinical AI: Areas of interest

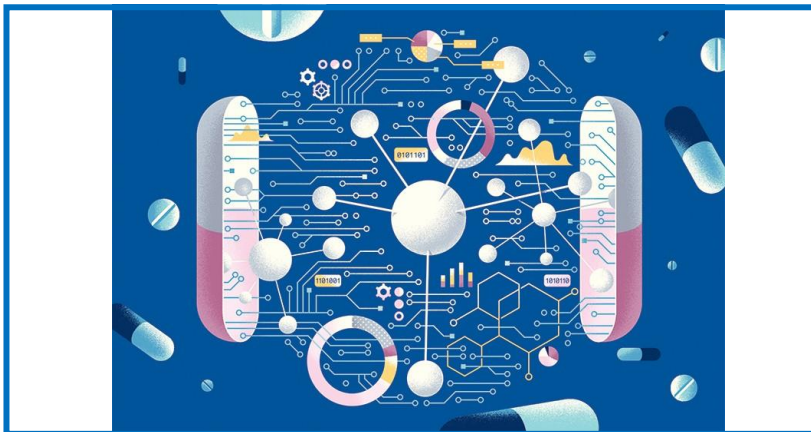
Computer vision



Natural language processing



Drug discovery



Electronic health records

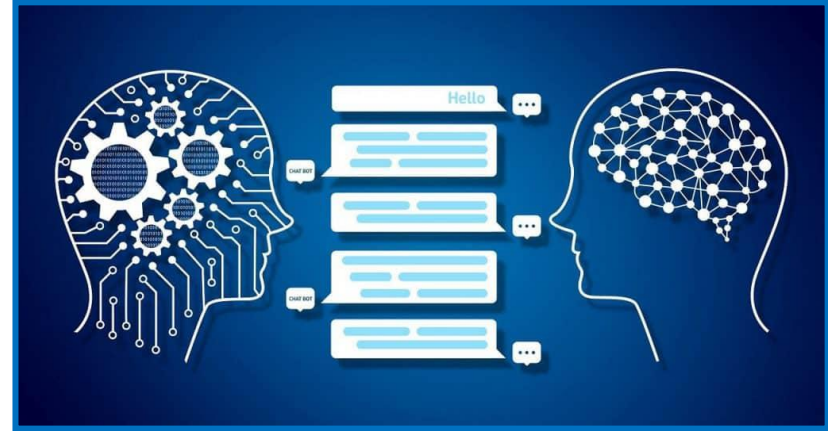


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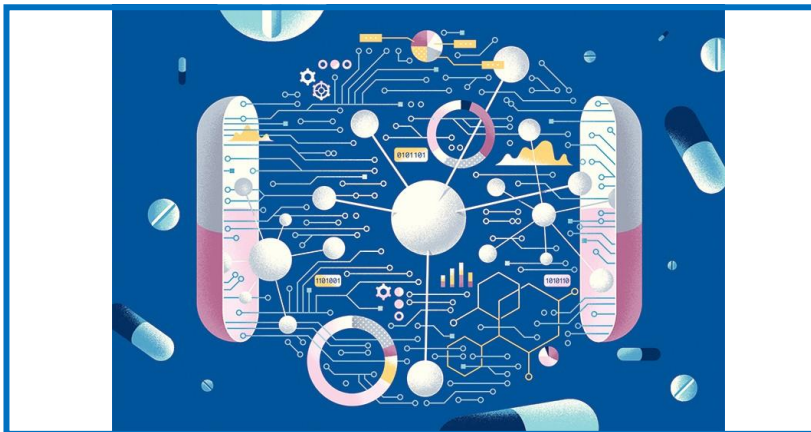
Computer vision

- Radiology
- Pathology
- Ophthalmology

Natural language processing



Drug discovery

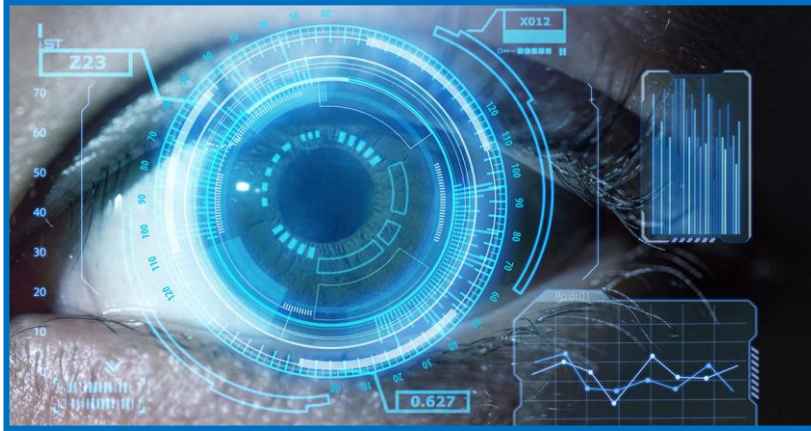


Electronic health records

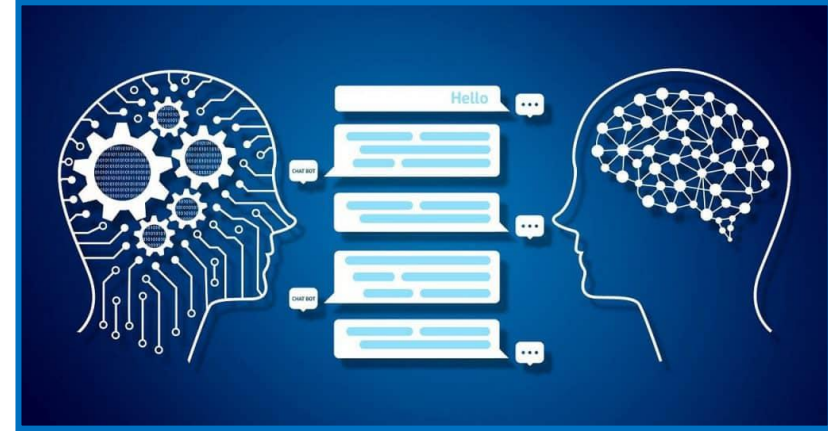


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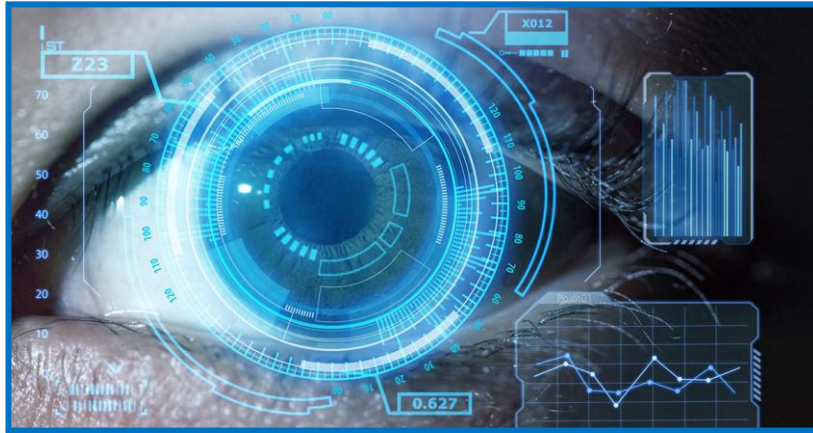
- Protein folding
- Molecular analysis
- Literature reviews

Electronic health records

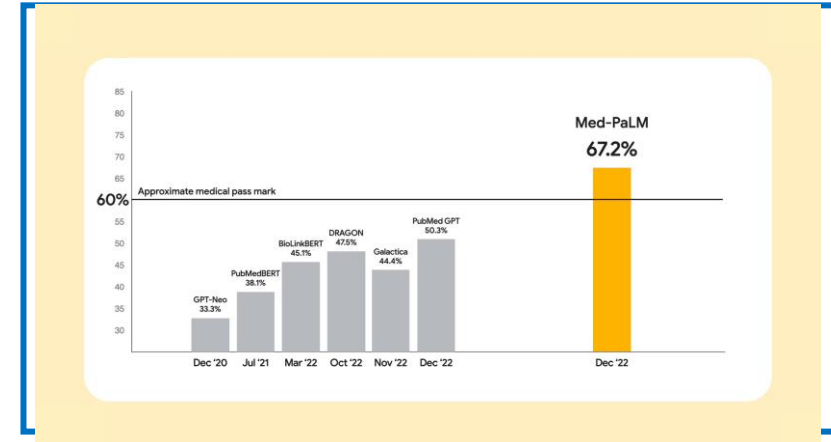


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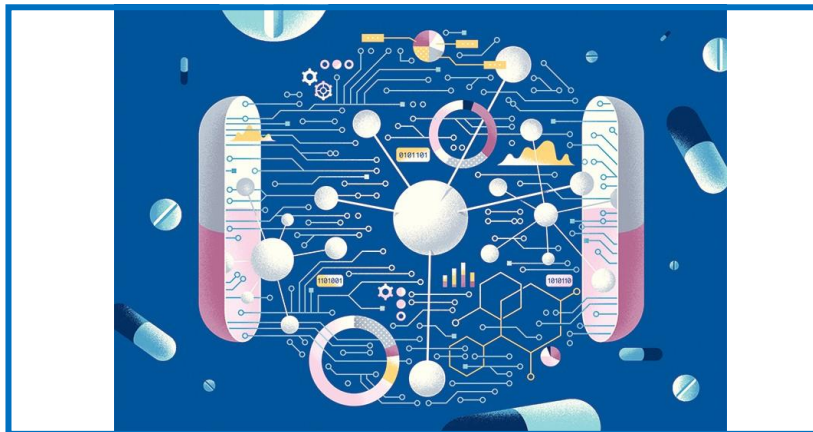
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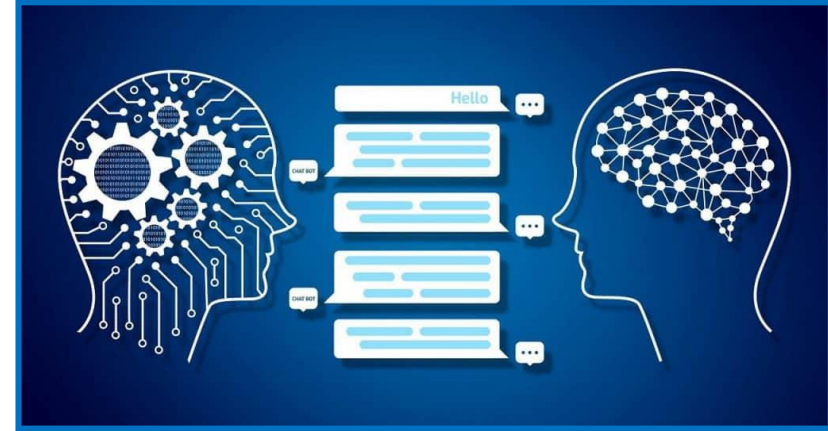


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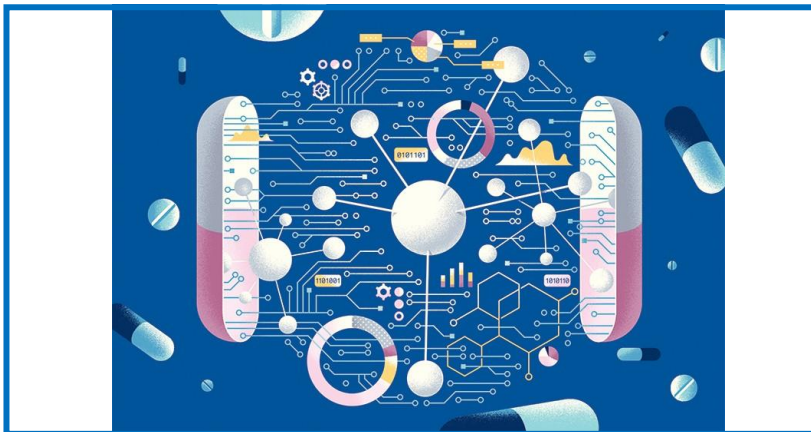
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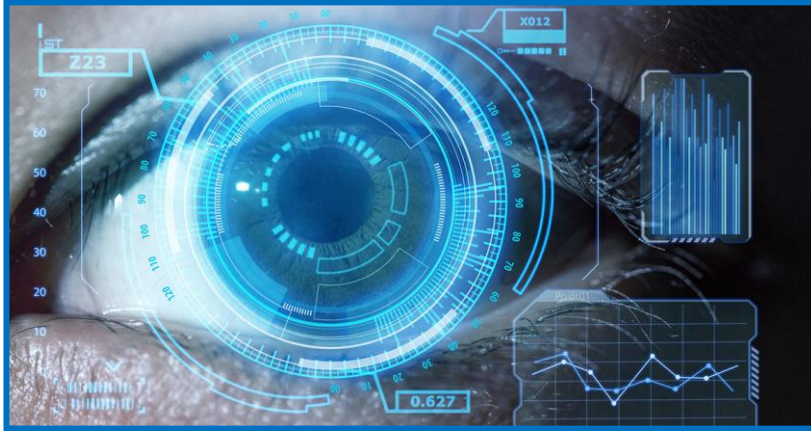


Electronic health records

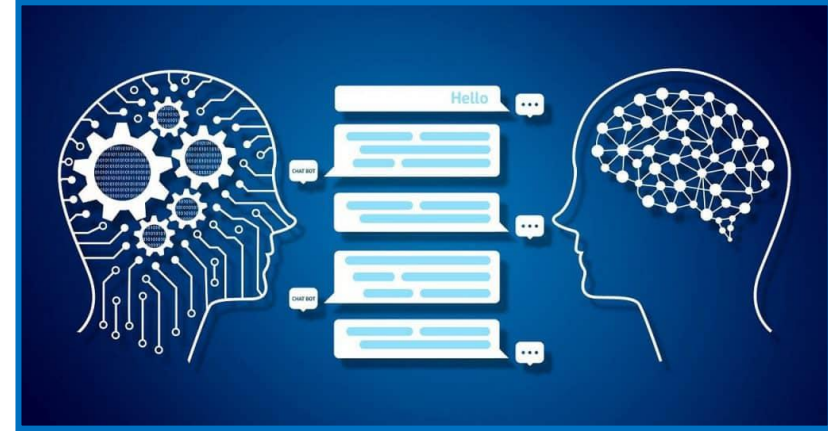
- Introducing structure
- Predicting outcomes

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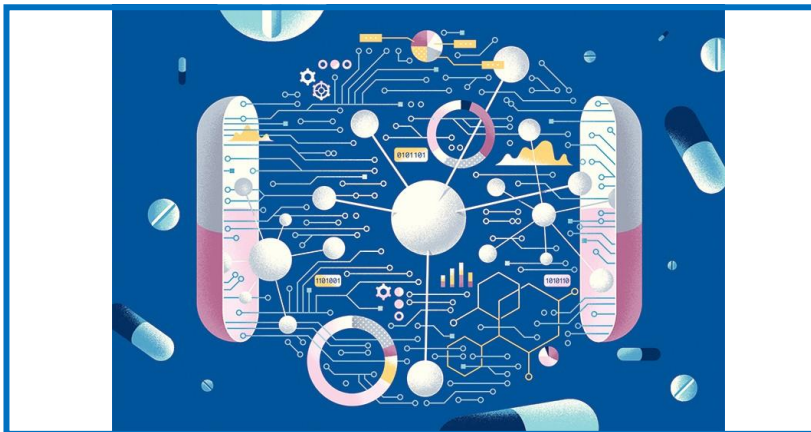
Computer vision



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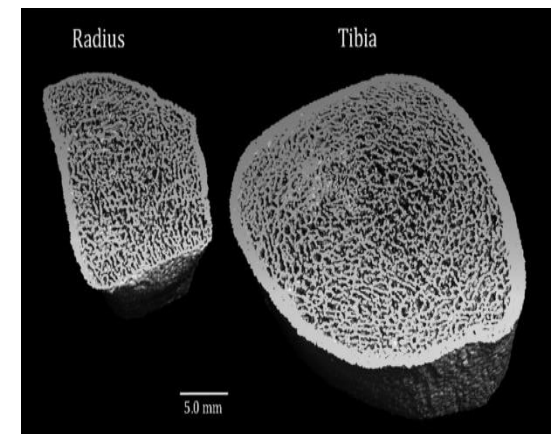
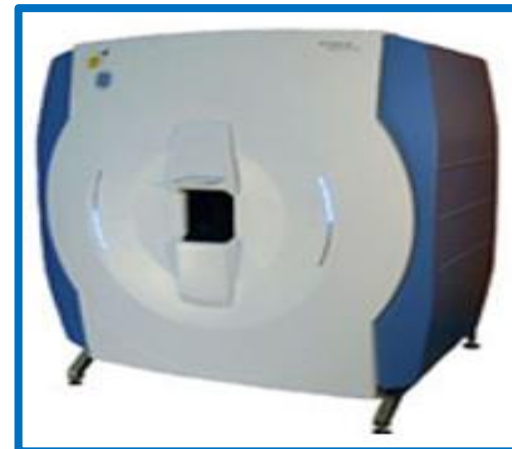
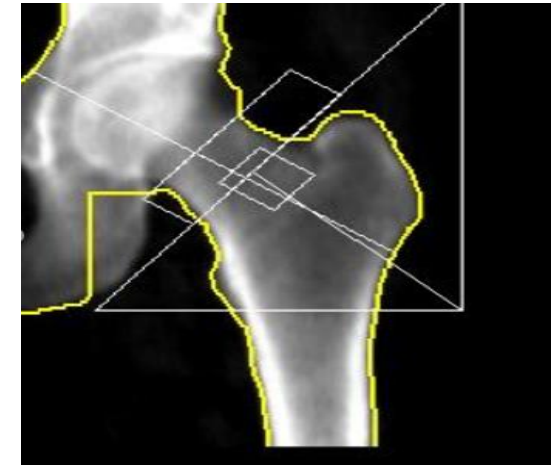


Electronic health records

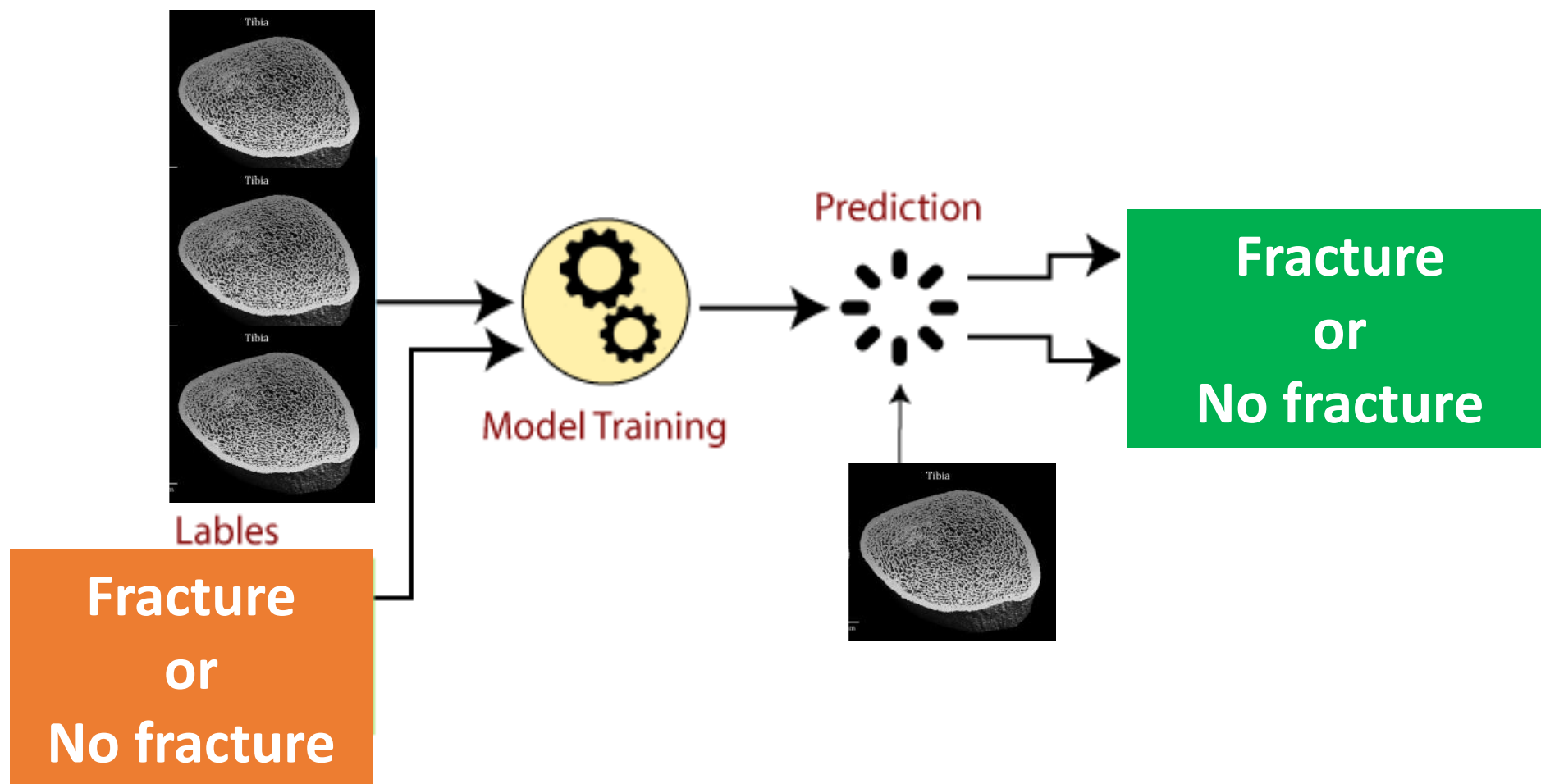


Background

- Osteoporosis
 - Highly prevalent
 - Difficult to identify
 - Current imaging practice: DXA
- Bone microarchitecture:
 - **HR-pQCT** (High-resolution peripheral quantitative computed tomography)
 - “Virtual bone biopsy”
 - Plethora of parameters
- More user-friendly method?

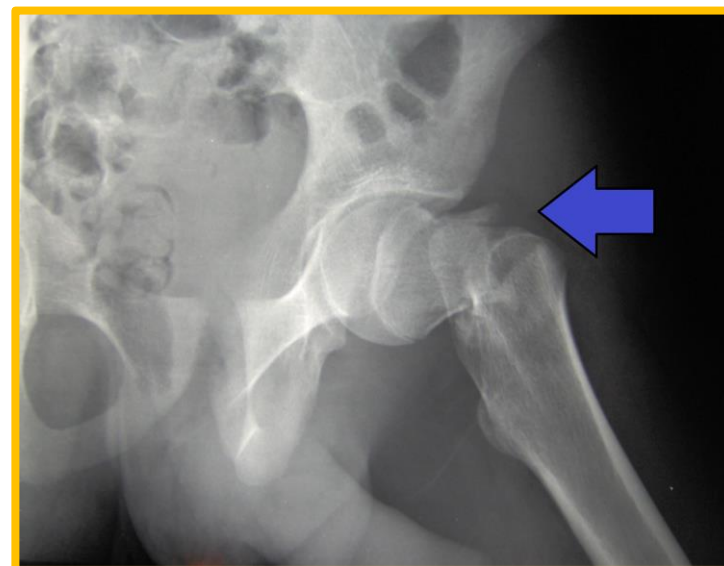
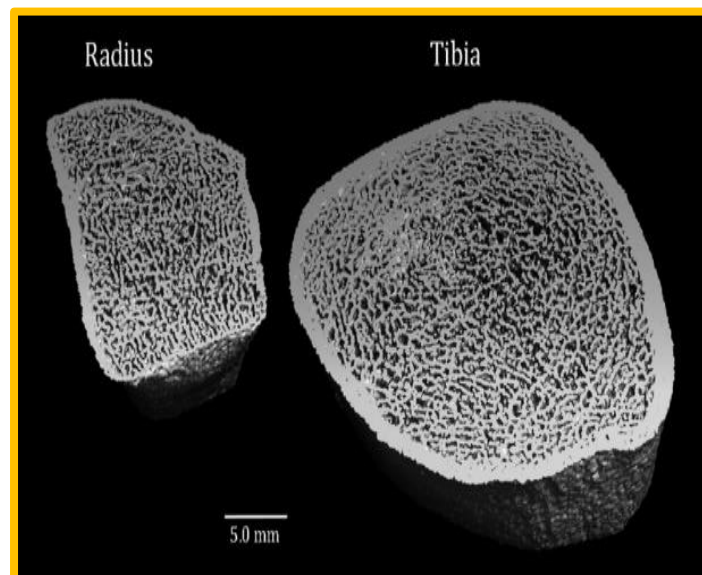


Supervised Machine Learning

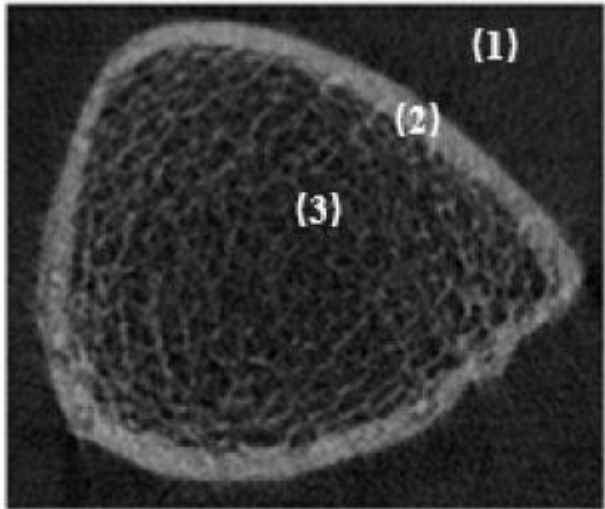


Objective

- Can computer vision and machine learning be to segment bone compartments
- Does this approach add to fracture discrimination?

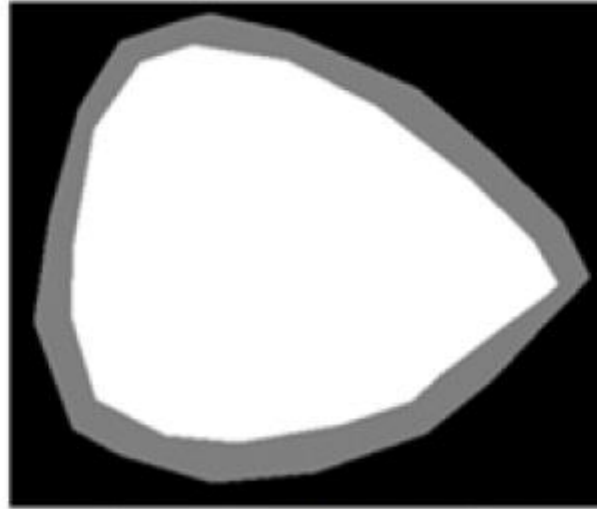


Automatic segmentation



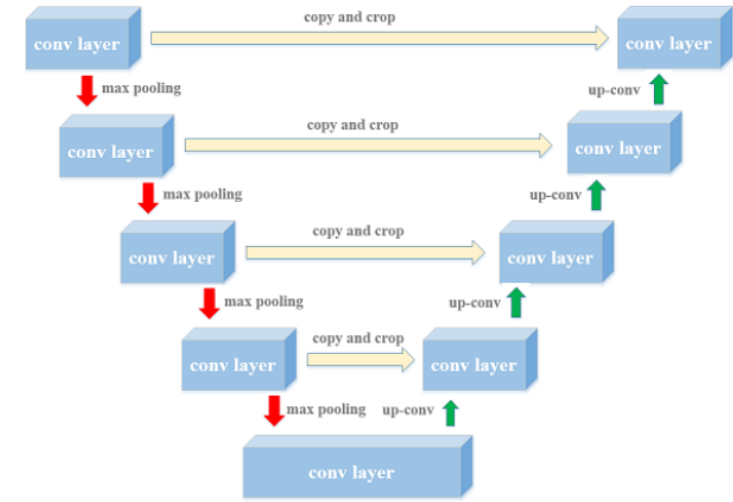
(a)

Original



(b)

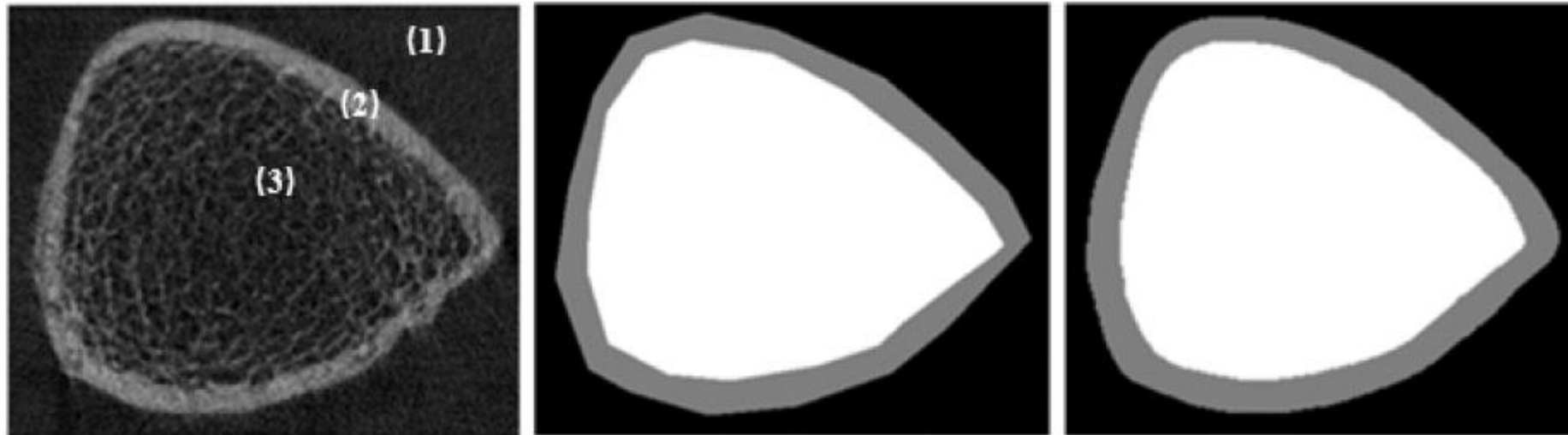
Manual
(47 scans)



(c)

Automatic
(U-Net)

Automatic segmentation



(a)

Original

(b)

Manual
(47 scans)

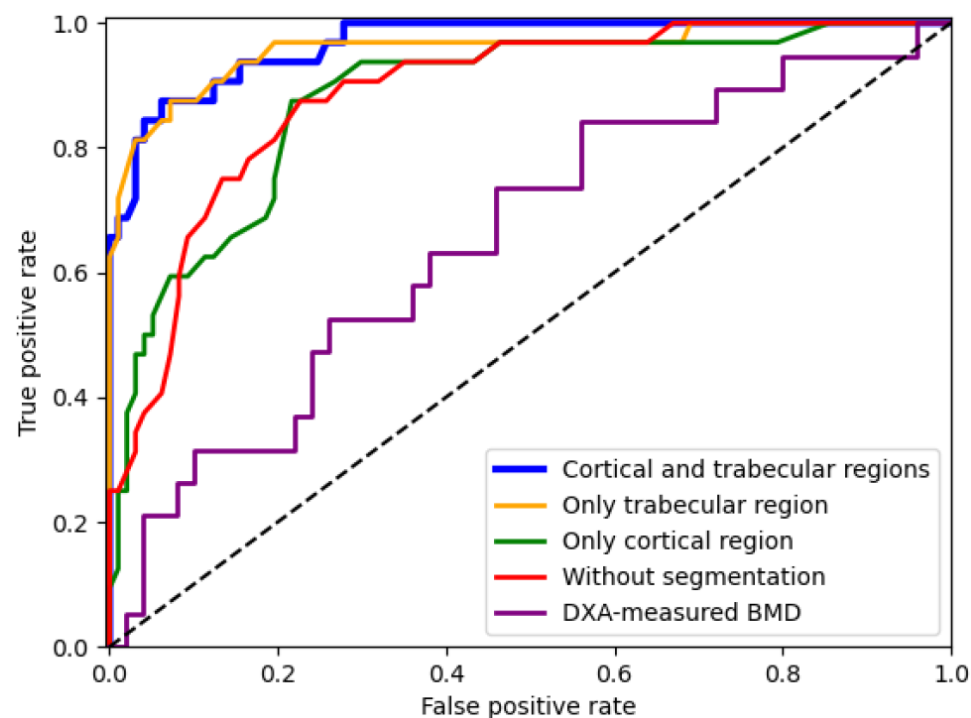
(c)

Automatic
(U-Net)

Intersection over union (IoU) = 96.5%

Automatic segmentation: Results

Input data	AUC (95% CI)	Sensitivity	Specificity
Cortical	0.88 (0.81-0.95)	0.72	0.80
Trabecular	0.95 (0.91-1.00)	0.94	0.84
Cortical and trabecular	0.97 (0.94-1.00)	0.88	0.87



Conclusion

- Deep learning method for performing 3D segmentation is effective
- Segmentation and feature fusion of cortical and trabecular compartments has greater association with fracture



Clinical **AI** Interest Group

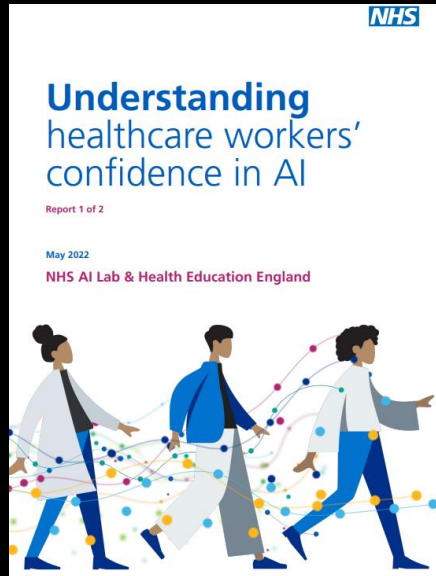
The Alan Turing Institute



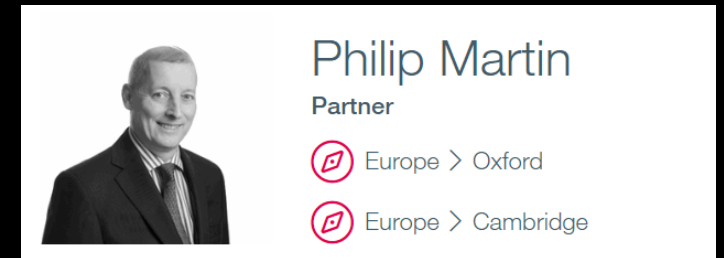
Sharing



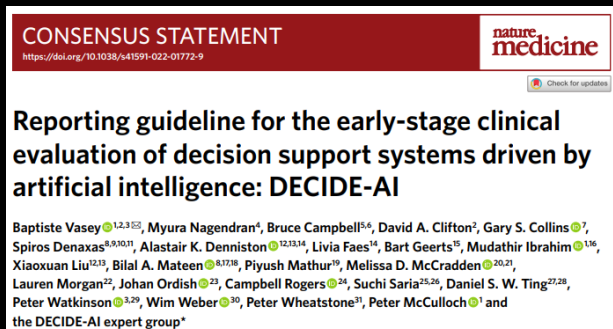
Sharing



Health Equity



Marks & Clark



Dom Cushnan
NHS AI Lab

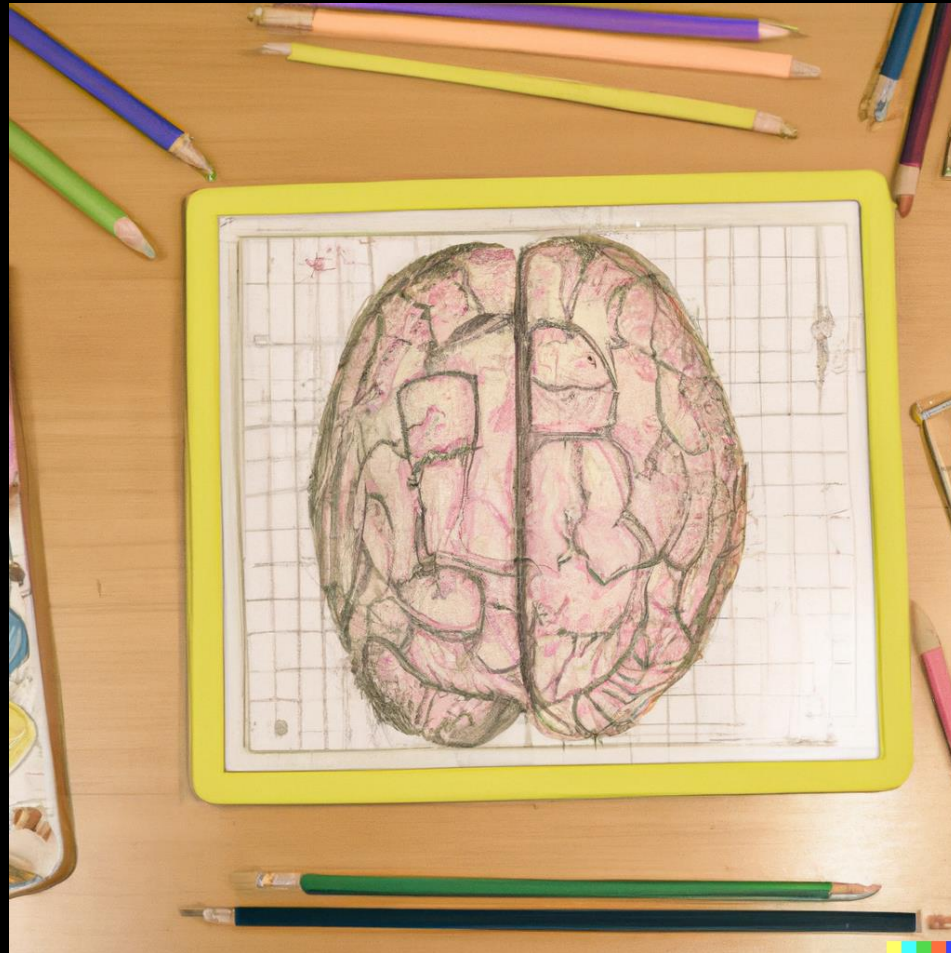


Alastair Denniston
RHC Report



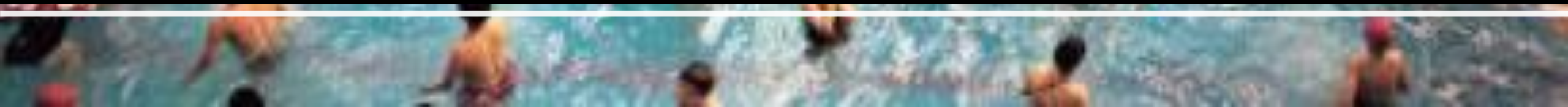
Educating

Summer School





Pooling



Clinical **AI** Interest Group

The Alan Turing Institute

Next meeting: Wednesday 21st June, 12pm