

# Data-driven & Model-driven Methods and their use in computer vision

Xiaohao Cai

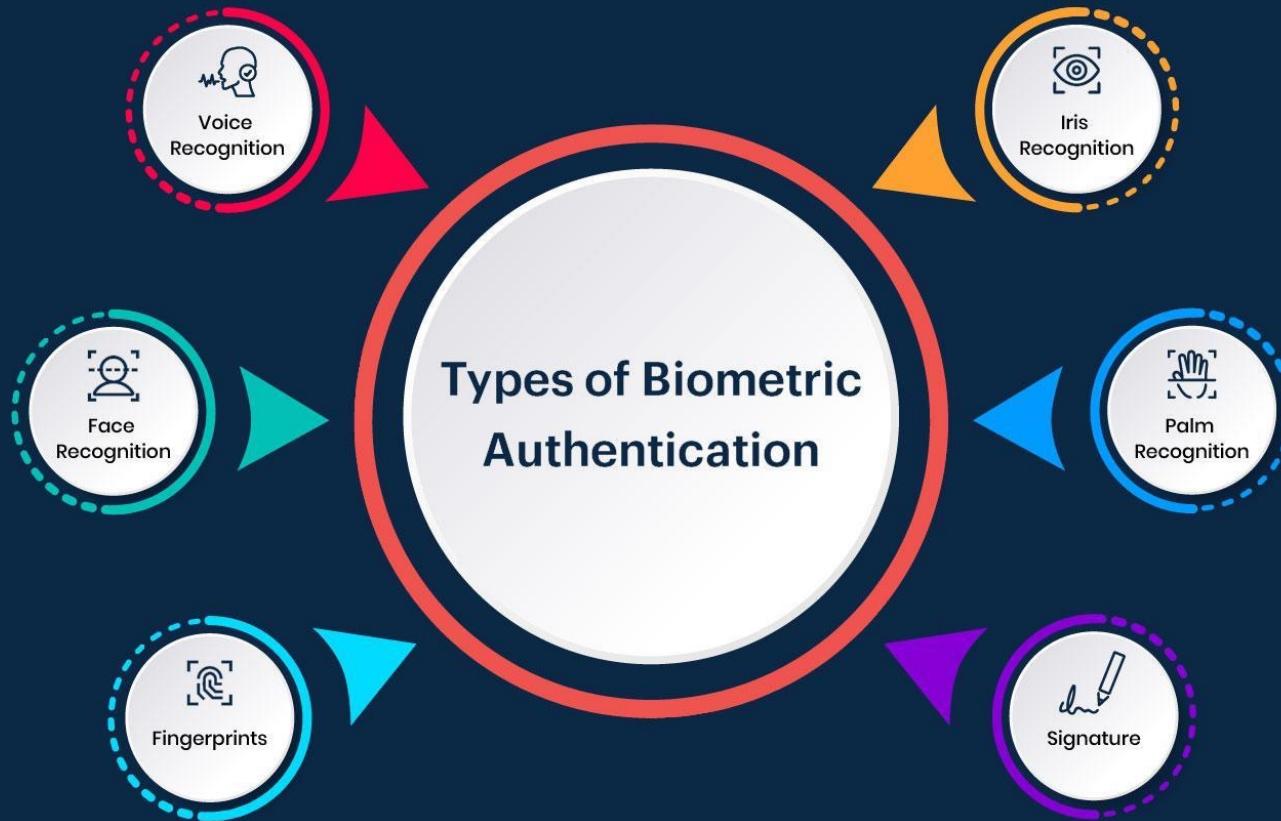
University of Southampton UK

COMP3204 Computer Vision

## What are their pros and cons?

# Content

1. Biometrics account for a large portion  
in computer vision
2. Some data-driven and model-driven methods  
in computer vision



# Different Types of Biometrics



Typing Style



Navigation Style

## Behavioral Biometric Identifiers



Interaction Style



Physical Style



Signature



Face Recognition



Eye Scanners



Fingerprints



Voice Recognition



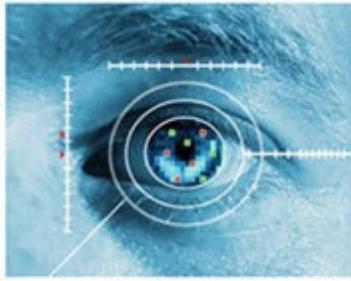
DNA

## Chemical & Vein Biometric Identifiers



Vein Recognition

Physiological



Iris



Fingerprint



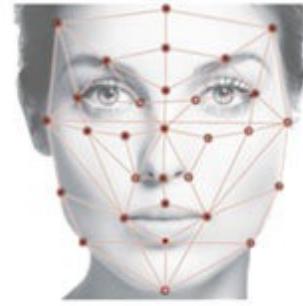
Ear



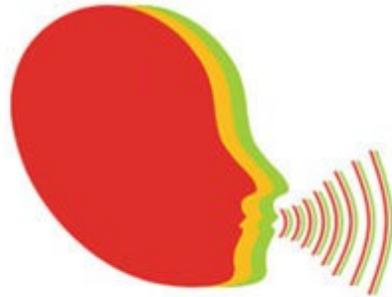
DNA



Vein print



Face



Voice



Gait



Signature

# Gait biometrics

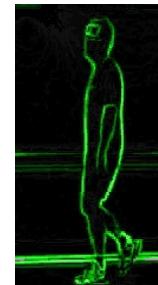
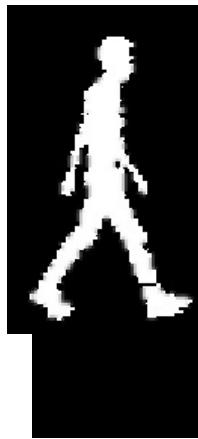
Identifying people  
by their gait



As a biometric, **gait** is available at a distance when other biometrics  
are obscured or at too **low resolution**

# Many gait representations possible

Recognising people from the motion of the **whole** body



silhouette

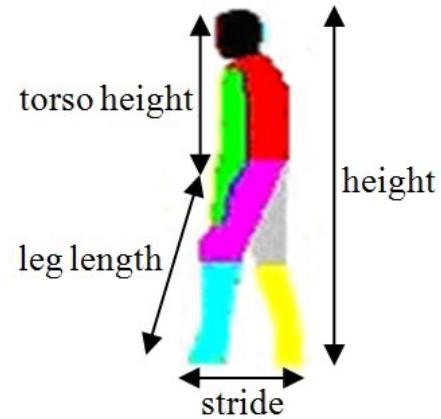
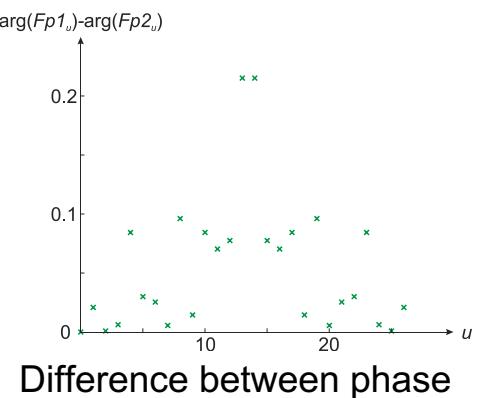
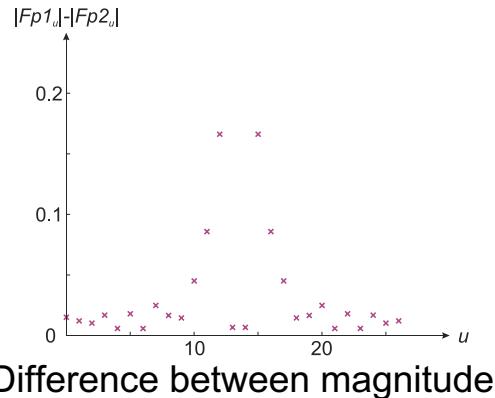
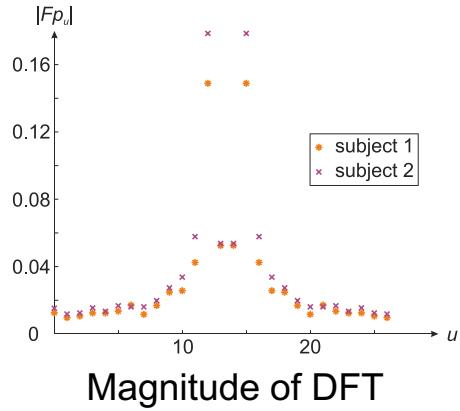
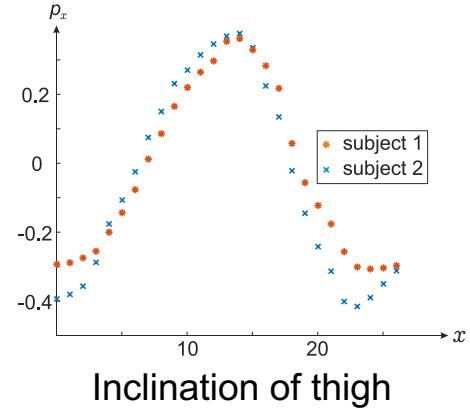
flow

edges

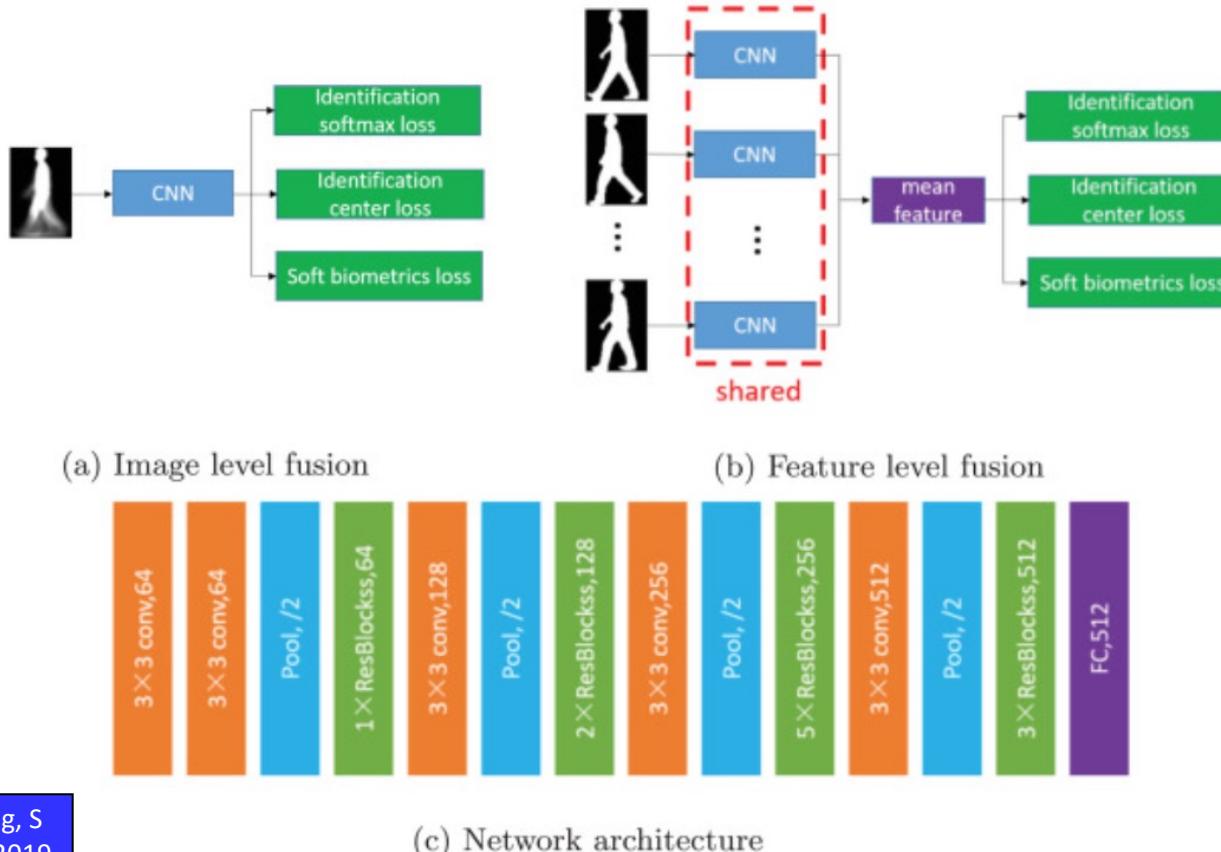
symmetry

acceleration

# Model-based recognition



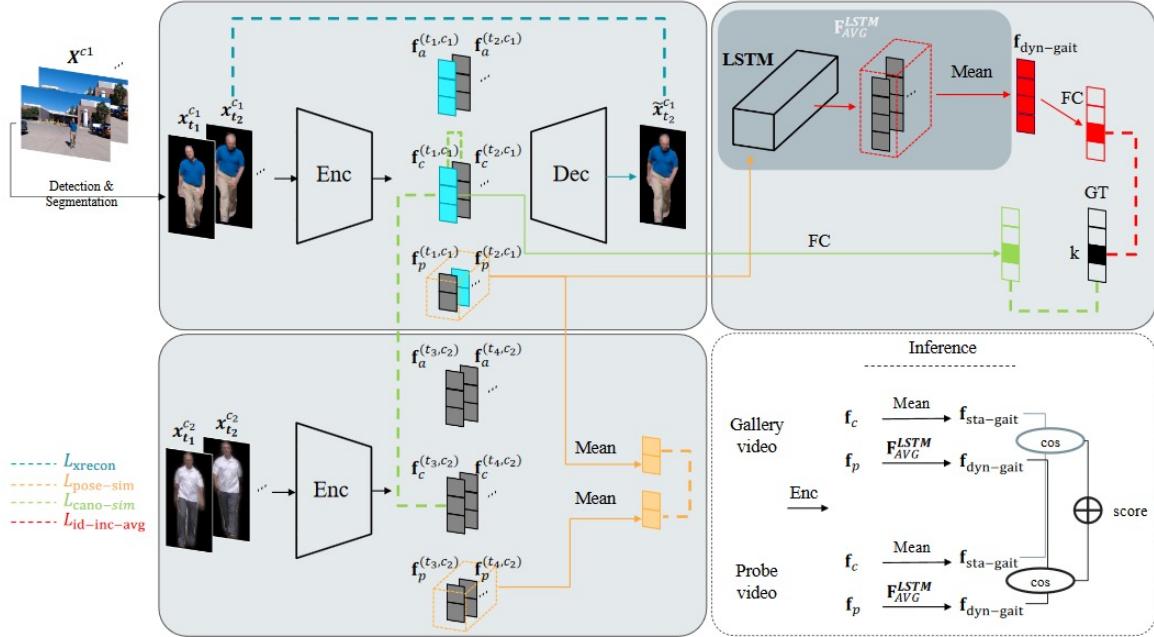
# Hand crafted then; deep learning now



# Recent works - Gait

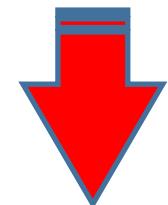
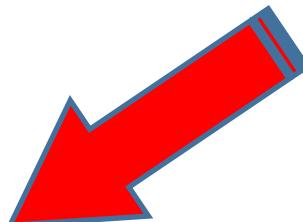
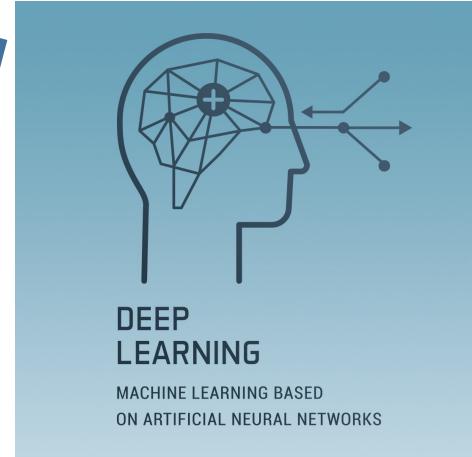
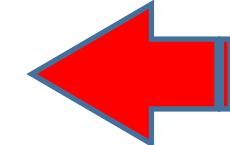
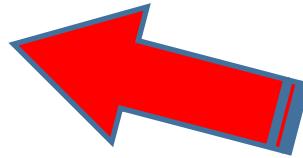
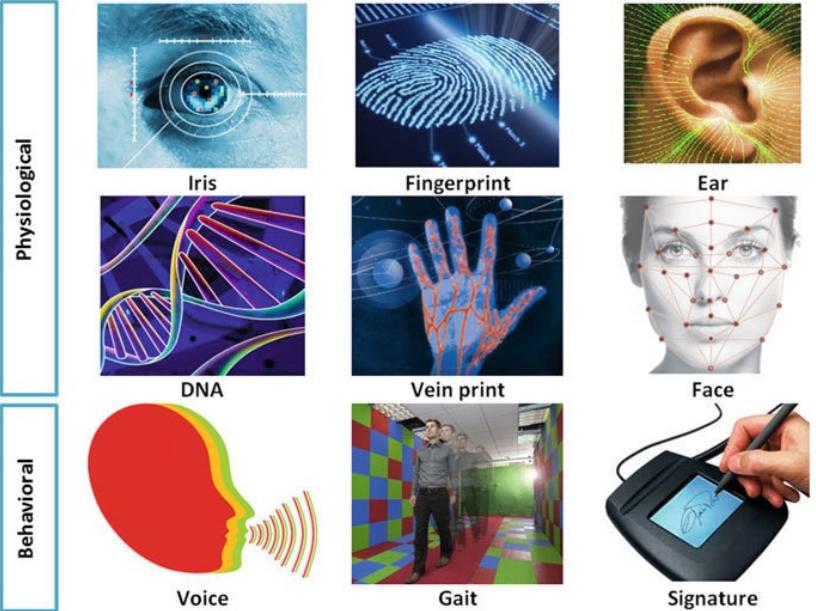


**Fig. 1.** Samples from the KinGaitWild dataset



SE Bekhouche, A Chergui, A Hadid...  
ICIP 2020

Z Zhang, L Tran, F Liu , X Liu,  
IEEE TPAMI 2019



**and many  
more ...**

IMAGE COLORING



IMAGE NOISE REDUCTION





**“DATA IS THE NEW GOLD”**

# Gait biometrics databases

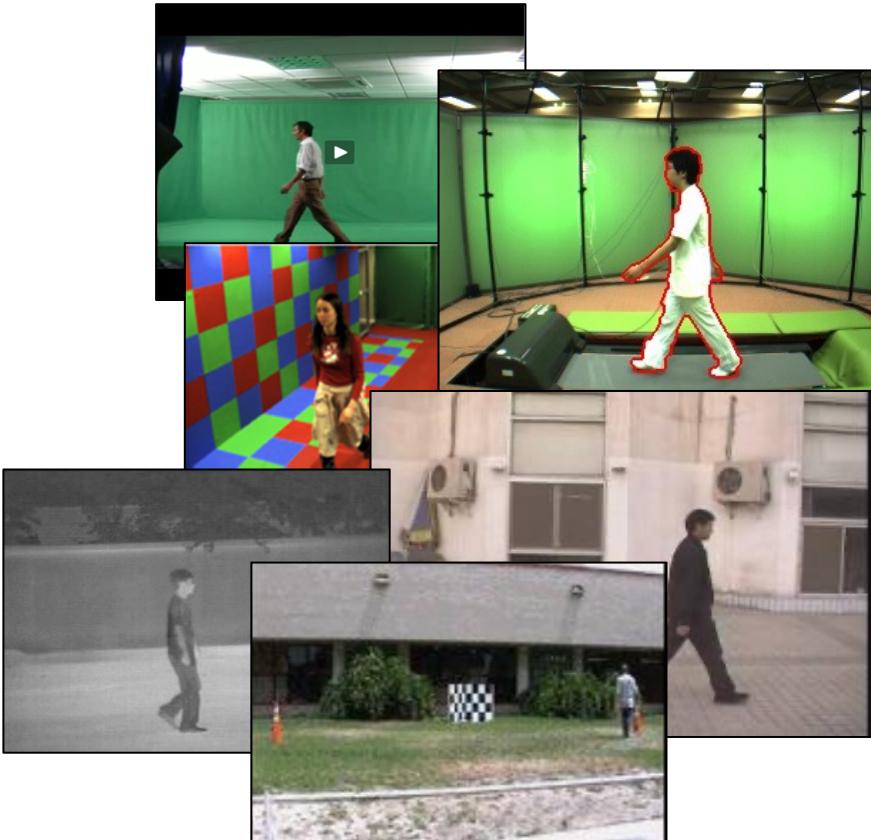
## Laboratory

- Southampton 3D and 2D
- CASIA (+ multiview, thermal)
- Osaka OU-ISIR (+ multiview)

## 'Real' World

- HumanID
- Southampton
- CASIA

+ accelerometer, footfall, medical



# What changes regarding datasets?

Many covariates can affect walking style

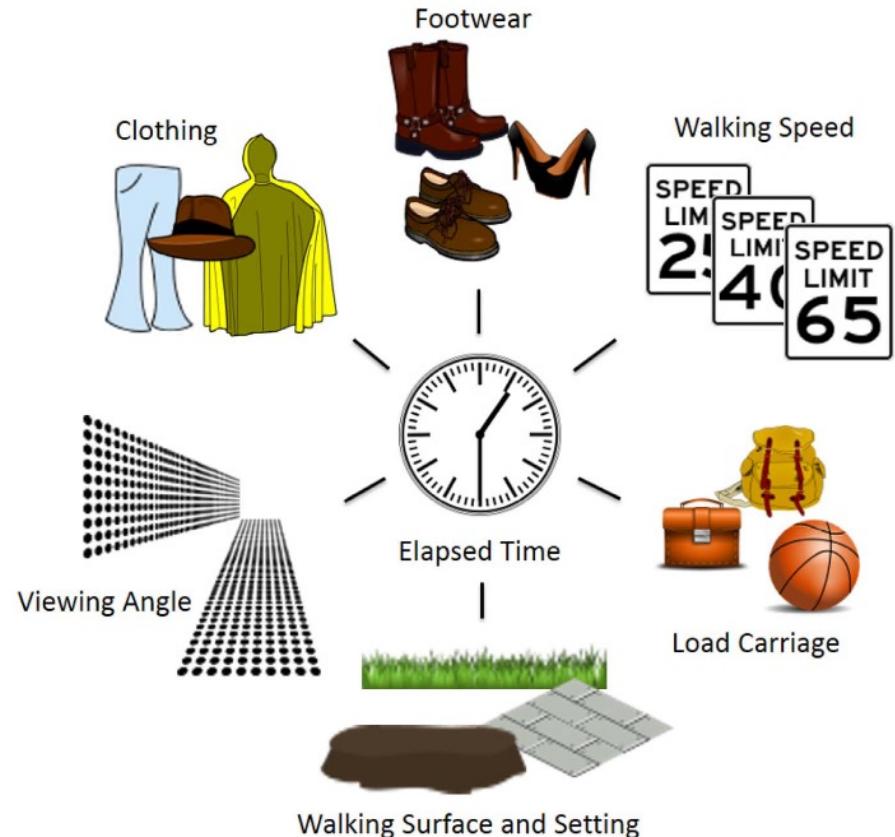
.... + health, drugs, mood ...

Domain shift

Class imbalance

Noisy annotation

*and more ...*





KEEP  
CALM  
AND  
USE DATA  
WISELY

## A Microsoft AI tool is helping to speed up cancer treatment – and Addenbrooke's will be the first hospital in the world to use it

December 9, 2020 | Microsoft reporter



<https://news.microsoft.com/en-gb/2020/12/09/a-microsoft-ai-tool-is-helping-to-speed-up-cancer-treatment-and-addenbrookes-will-be-the-first-hospital-in-the-world-to-use-it/>

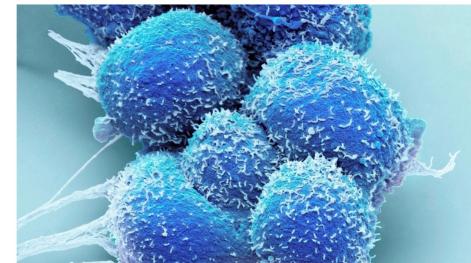
# Medical imaging with AI

THE TIMES Today's sections ▾ Past six days Explore ▾ Times Radio Log in

## Inner Eye AI identifies tumours to speed up treatment of cancer

Katie Gibbons

Monday January 11 2021, 12.01am,  
The Times



The Inner Eye software is the result of an eight-year project with Microsoft and Addenbrooke's hospital ALAMY

A hospital in Cambridge is the first to use artificial intelligence technology developed by Microsoft to treat cancer patients faster, helping to cut the treatment backlog and save lives.

## NEWS

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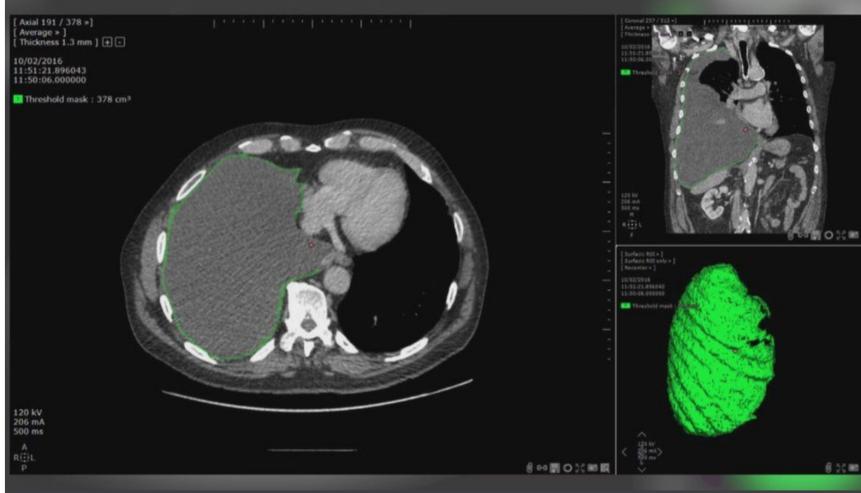
## AI technology used to track asbestos cancer tumours

By Laura Goodwin  
BBC Scotland Innovations Correspondent

① 2 days ago



# Medical imaging with AI



**Research****Use of artificial  
programmes:**

BMJ 2021 ;374 :d5162  
Cite this as: BMJ 2021;  
374:d5162

[Article](#)[Related](#)[Karoline Freeman](#)[Daniel Todkill](#)[Aileen Clarke, professor](#)**Author affiliation**

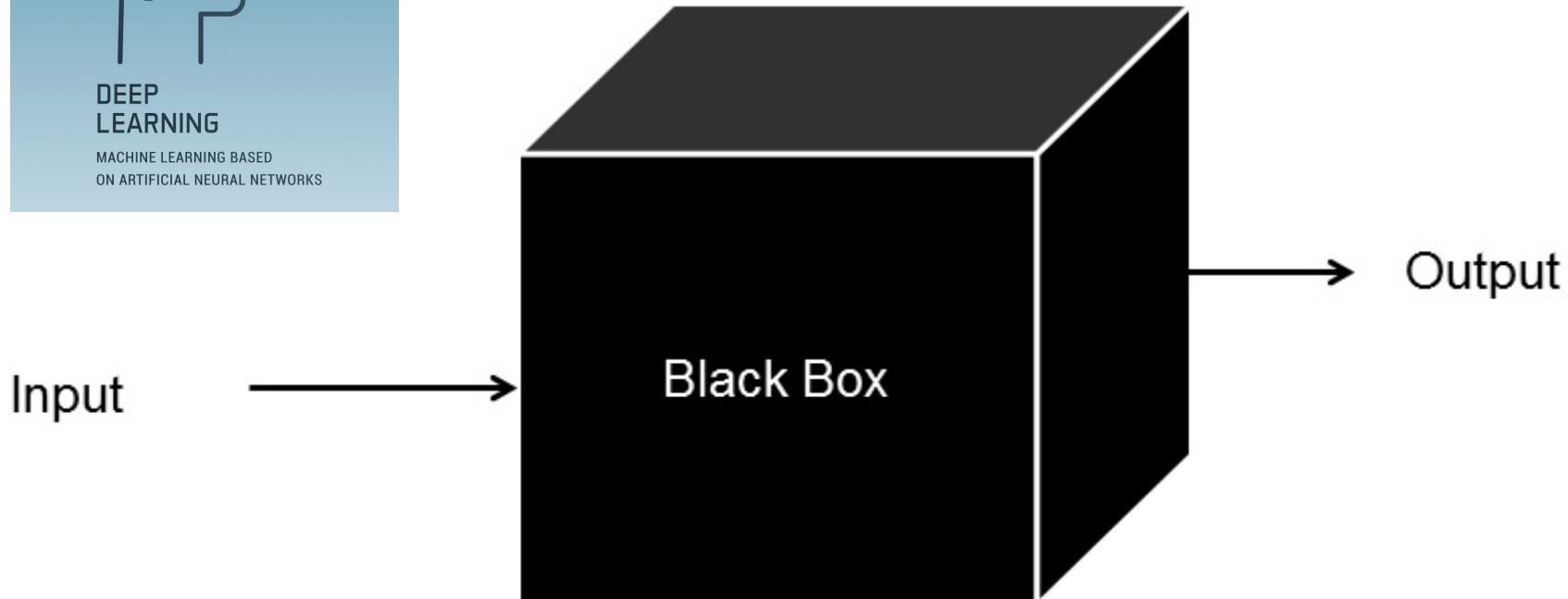
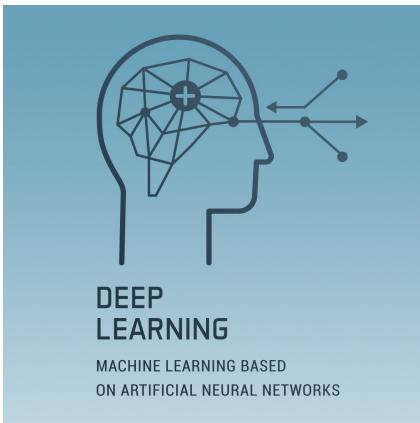
Correspondence to

Accepted 21 July 2021

[Abstract](#)

## Conclusions

Current evidence on the use of AI systems in breast cancer screening is a long way from having the quality and quantity required for its implementation into clinical practice. Well designed comparative test accuracy studies, randomised controlled trials, and cohort studies in large screening populations are needed which evaluate commercially available AI systems in combination with radiologists. Such studies will enable an



*Internal behavior of the code is unknown*

## What it means to look inside the black box

**Explainability**

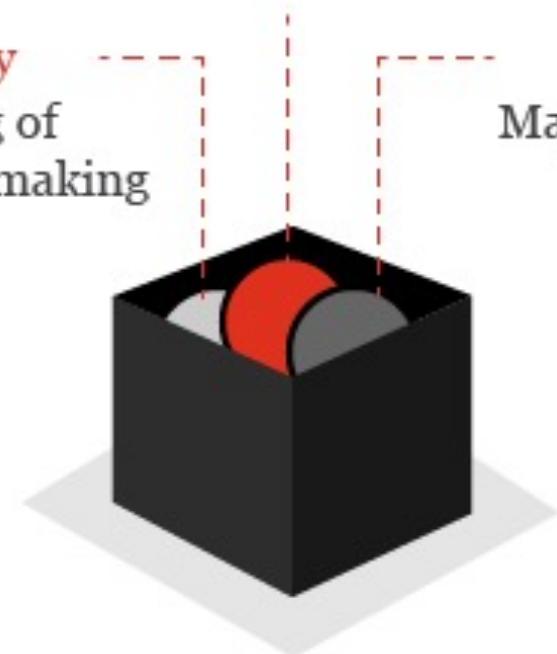
Understanding reasoning  
behind each decision

**Transparency**

Understanding of  
AI model decision making

**Provability**

Mathematical certainty  
behind decisions



# POST

[UK Parliament](#) > [POST](#) > [Interpretable machine learning](#)

Research Briefing

## Interpretable machine learning

Published Tuesday, 06 October, 2020

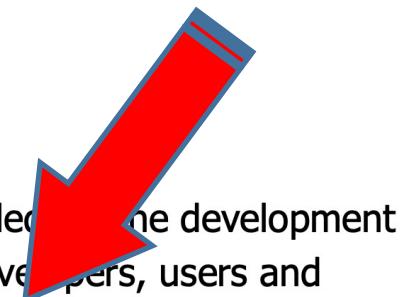
[POSTnote](#) [Crime and justice](#) [Digital tech](#) [Health and social care](#) [Transport and infrastructure](#) [Research](#)

✉ [Lorna Christie](#)

Machine learning (ML, a type of artificial intelligence) is increasingly being used to support decision making in a variety of applications including recruitment and clinical diagnoses. While ML has many advantages, there are

concerns that in some cases it may not be possible to explain or understand how a complex ML system has reached its output, making ML easier to interpret. It also gives a brief overview of systems more accountable.

In 2018, the Lords Committee on AI called for the development of AI systems that are “intelligible to developers, users and regulators”. It recommended that an AI system that could have a substantial impact on an individual’s life should not be used unless it can produce an explanation of its decisions.<sup>4</sup> In a

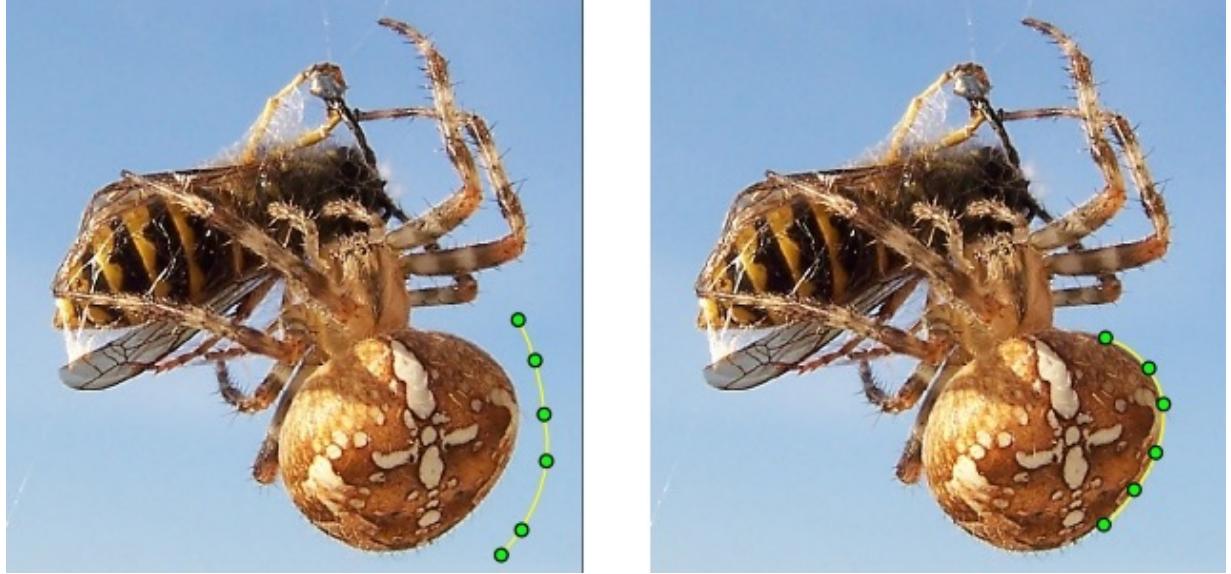




KEEP  
CALM  
AND  
USE DATA  
WISELY

# Model-driven Methods

## Active Contours

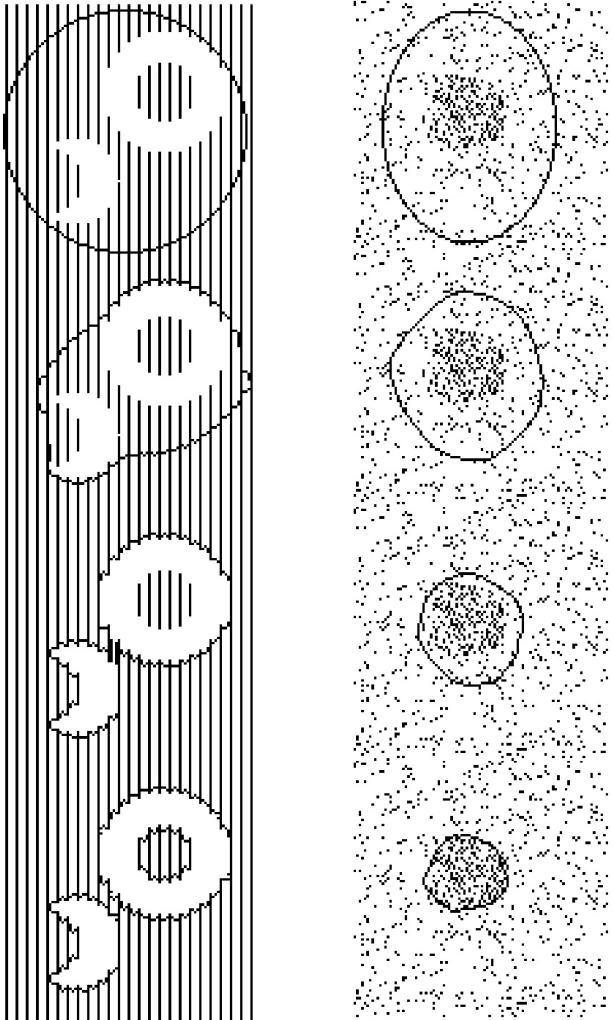


International Journal of Computer Vision, 321–331 (1988)  
© 1987 Kluwer Academic Publishers, Boston, Manufactured in The Netherlands

### Snakes: Active Contour Models

MICHAEL KASS, ANDREW WITKIN, and DEMETRI TERZOPoulos  
*Schlumberger Palo Alto Research, 3340 Hillview Ave., Palo Alto, CA 94304*

[credit: wikipedia]



## Active Contours Without Edges

Tony F. Chan, *Member, IEEE*, and Luminita A. Vese

**Abstract**—In this paper, we propose a new model for active contours to detect objects in a given image, based on techniques of curve evolution, Mumford–Shah functional for segmentation and level sets. Our model can detect objects whose boundaries are not necessarily defined by gradient. We minimize an energy which can be seen as a particular case of the minimal partition problem. In the level set formulation, the problem becomes a “mean-curvature flow”-like evolving the active contour, which will stop on the desired boundary. However, the stopping term does not depend on the gradient of the image, as in the classical active contour models, but is instead related to a particular segmentation of the image. We will give a numerical algorithm using finite differences. Finally, we will present various experimental results and in particular some

the image (the external energy). Observe that, by minimizing the energy (1), we are trying to locate the curve at the points of maxima  $|\nabla u_0|$ , acting as an edge-detector, while keeping a smoothness in the curve (object boundary).

A general edge-detector can be defined by a positive and decreasing function  $g$ , depending on the gradient of the image  $u_0$ , such that

$$\lim_{z \rightarrow \infty} g(z) = 0.$$

For instance

# Models

Models proposed in our work, e.g.:

- ▶  $\min_x \left\{ \frac{\lambda}{2} \|y - \mathcal{A}x\|_2^2 + \|\mathcal{W}x\|_1 \right\}$
- ▶  $\min_g \left\{ \frac{\lambda}{2} \|f - \mathcal{A}g\|_2^2 + \frac{\mu}{2} \|\nabla g\|_2^2 + \|\nabla g\|_1 \right\}$
- ▶  $\mu\Phi(f, \mathcal{A}g) + \lambda\Psi(g, u_i, c_i) + \sum_{i=1}^K \int_{\Omega} |\nabla u_i|$   
s.t.  $\sum_{i=1}^K u_i(x) = 1, u_i(x) \in \{0, 1\}$
- ▶  $\min_{u \in S} \left\{ \frac{1}{2} \|f - \mathcal{B}u\|_2^2 + \lambda \|\nabla u\|_0 \right\}$
- ▶  $\min_{\psi} \left\{ D[T(\psi), R] + \alpha \|\Delta\psi\|_2^2 \right\}$

Convex optimisation  
algorithms

- ▶ ADMM
- ▶ Primal-dual
- ▶ Split-Bregman
- ▶ Augmented  
Lagrangian

Sparse regularizations

- ▶  $\|\cdot\|_0, \|\cdot\|_1, \|\cdot\|_2$
- ▶ with  $\nabla, \Delta, \mathcal{W}$
- ▶  $\mathcal{W}$ : Wavelet transform

# T-ROF (*Thresholded-ROF*)

[SISC, '19; EMMCVPR, '13]

X. Cai, R. Chan, C.-B. Schönlieb  
G. Steidl, T. Zeng

## Image Restoration

ROF model  
(1992, citation > 15,700)

thresholding

## Image Segmentation

Chan-Vese model  
(2001, citation > 12,600)

$$\min_{u \in BV(\Omega)} \left\{ TV(u) + \frac{\mu}{2} \int_{\Omega} (f - u)^2 dx \right\},$$

$TV(u)$ : total variation of  $u$

$$\min_{\Omega_i; m_i} \left\{ \text{Per}(\Omega_1; \Omega) + \lambda \sum_{i=0}^1 \int_{\Omega_i} (m_i - f)^2 dx \right\},$$

$$\Omega := \Omega_0 + \Omega_1$$

$\text{Per}(\Omega_1; \Omega)$ : perimeter of set  $\Omega_1$

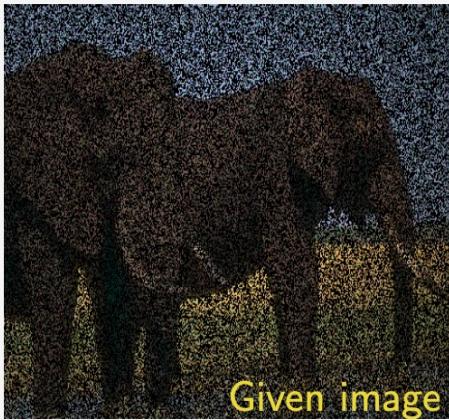
### Theorem

(Relation between ROF and Chan-Vese model) *Let  $u^* \in BV(\Omega)$  solve the ROF model. For given  $0 < m_0 < m_1 \leq 1$ , let  $\tilde{\Sigma} := \{x \in \Omega : u^*(x) > \frac{m_1+m_0}{2}\}$  fulfill  $0 < |\tilde{\Sigma}| < |\Omega|$ . Then  $\tilde{\Sigma}$  is a minimizer of the Chan-Vese model for  $\lambda := \frac{\mu}{2(m_1-m_0)}$  and fixed  $m_0, m_1$ . In particular,  $(\tilde{\Sigma}, m_0, m_1)$  is a partial minimizer of the Chan-Vese model if  $m_0 = \text{mean}_f(\Omega \setminus \tilde{\Sigma})$  and  $m_1 = \text{mean}_f(\tilde{\Sigma})$ .*

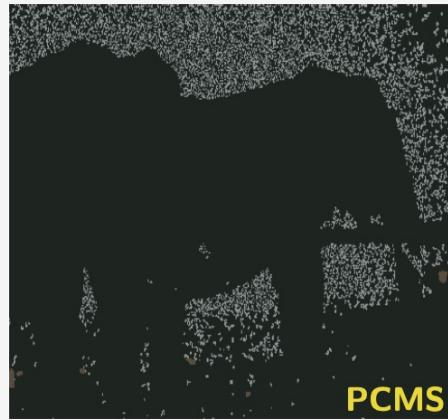
Colour image

**Research Grants Council  
of Hong Kong**  
香港 研究資助局

Method: SLaT



Given image



PCMS



Potts

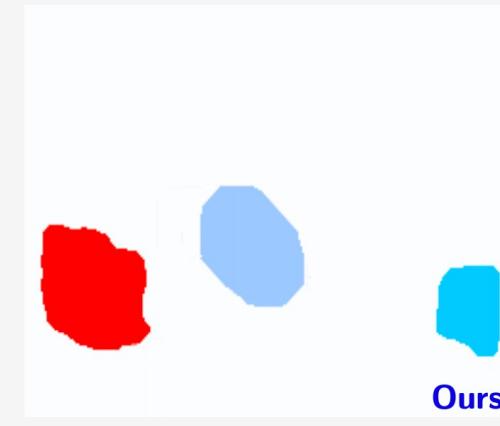


Ours

## Disparity and optical flow

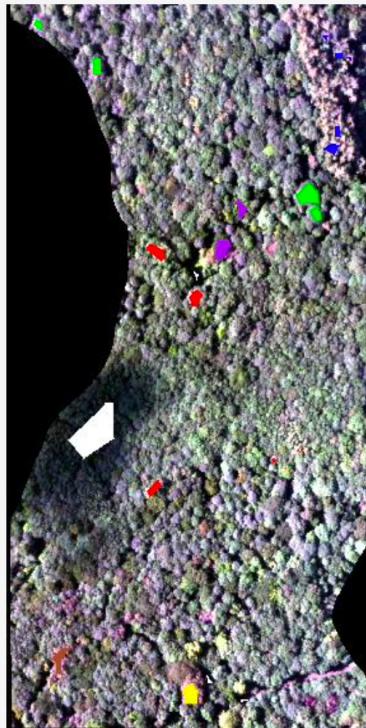
Algorithm for  
NP-hard problem

$$\|\nabla \cdot\|_0$$

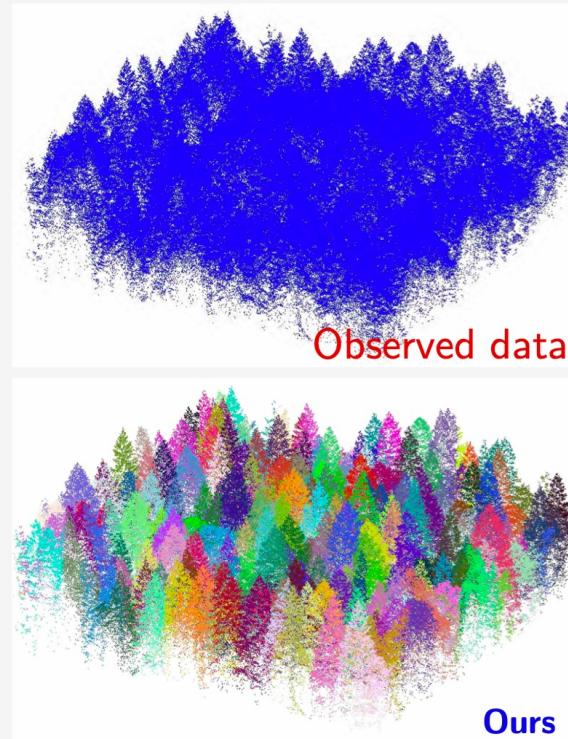


Computer  
vision

Hyperspectral



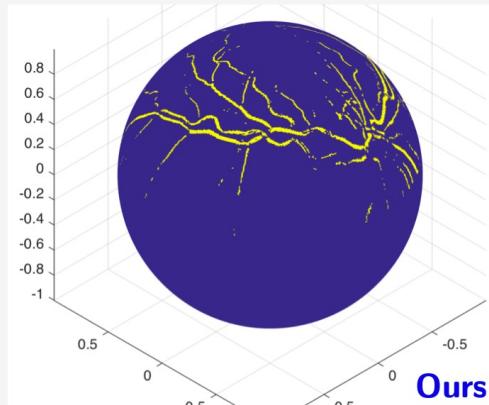
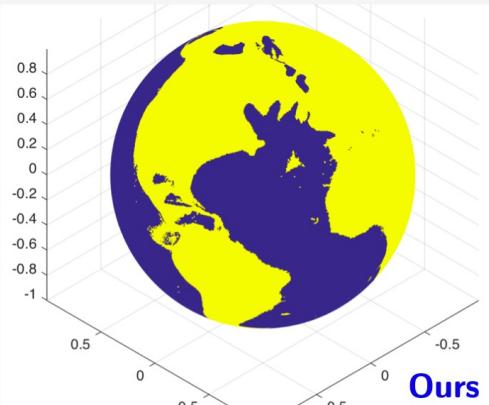
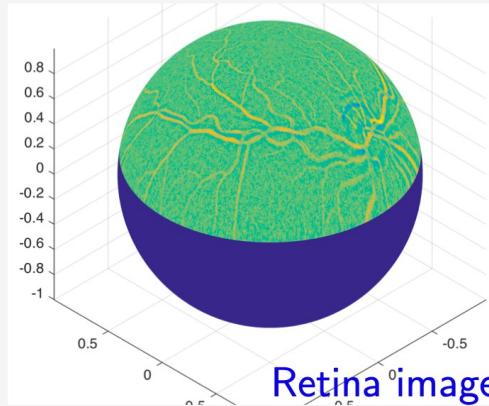
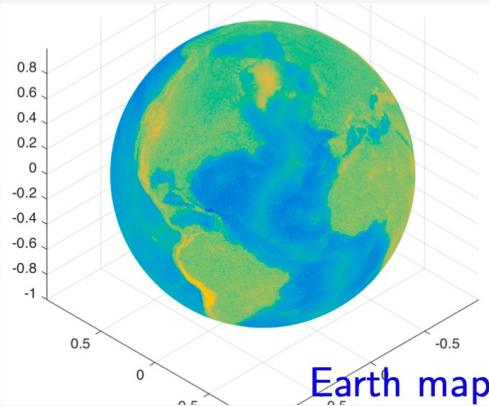
LiDAR



## Spherical image

Wavelet-based  
algorithm

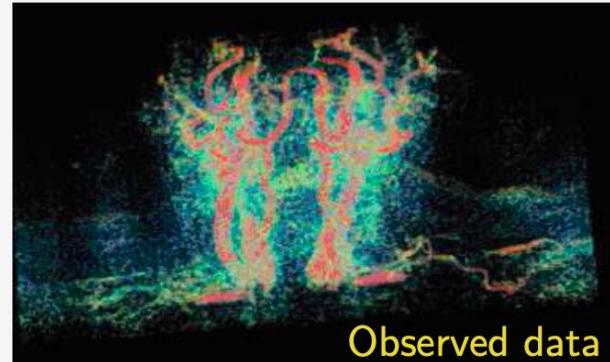
First  
segmentation paper  
on the sphere



SIIMS, '13  
SSVM, '12  
X. Cai, et al.

### 3D image – tubular

Wavelet-based  
algorithm



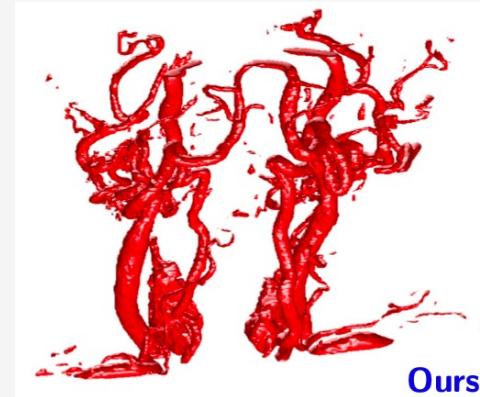
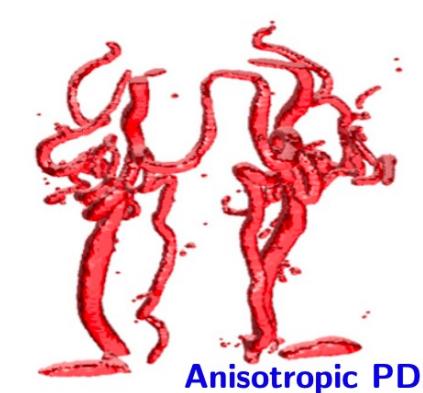
Data provided:

Prof. S. Morigi

Prof. F. Sgallari

Uni. of Bologna

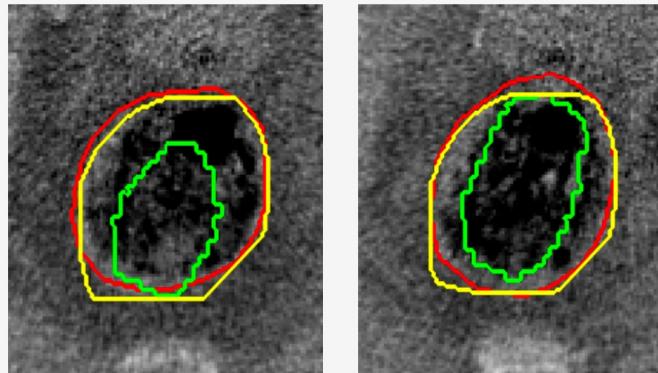
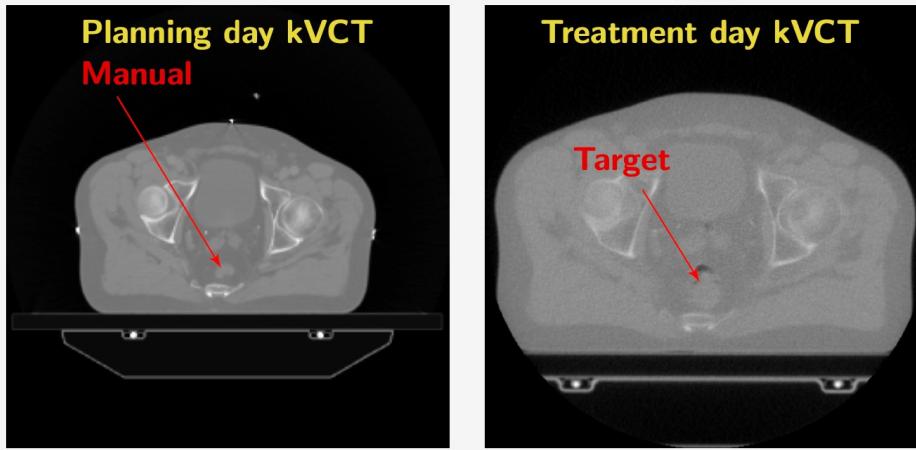
Italy



## Image-guided radiotherapy

CIJ, '17

EUSIPCO, '21



<http://www.componc.org/research/voxtor>

# Recent: models leveraging deep learning

Models proposed in our work, e.g.:

- ▶  $\min_x \left\{ \frac{\lambda}{2} \|y - \mathcal{A}x\|_2^2 + \|\mathcal{W}x\|_1 \right\}$
- ▶  $\min_g \left\{ \frac{\lambda}{2} \|f - \mathcal{A}g\|_2^2 + \frac{\mu}{2} \|\nabla g\|_2^2 + \|\nabla g\|_1 \right\}$
- ▶  $\mu\Phi(f, \mathcal{A}g) + \lambda\Psi(g, u_i, c_i) + \sum_{i=1}^K \int_{\Omega} |\nabla u_i|$   
s.t.  $\sum_{i=1}^K u_i(x) = 1, u_i(x) \in \{0, 1\}$
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- ▶  $\|\cdot\|_0, \|\cdot\|_1, \|\cdot\|_2$
- ▶ with  $\nabla, \Delta, \mathcal{W}$
- ▶  $\mathcal{W}$ : Wavelet transform

# Conclusions

1. Computer vision works and has a great future
2. Big difference between data-driven and model-driven
3. Gap is becoming smaller
4. What will happen in the future?

**We have more to learn ...**