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NETZEROAICT

Digital Contrast for Computerised Tomography

- Towards Climate Neutral and Sustainable Health Systems -

Towards **NetZero** Healthcare

**Mitigating the environmental impact of radiology
with novel AI imaging solutions**

Regent Lee

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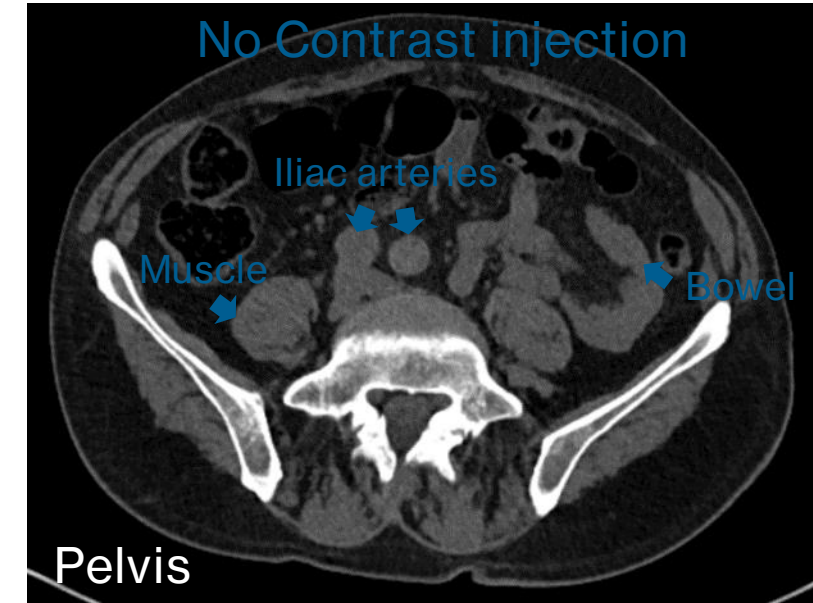
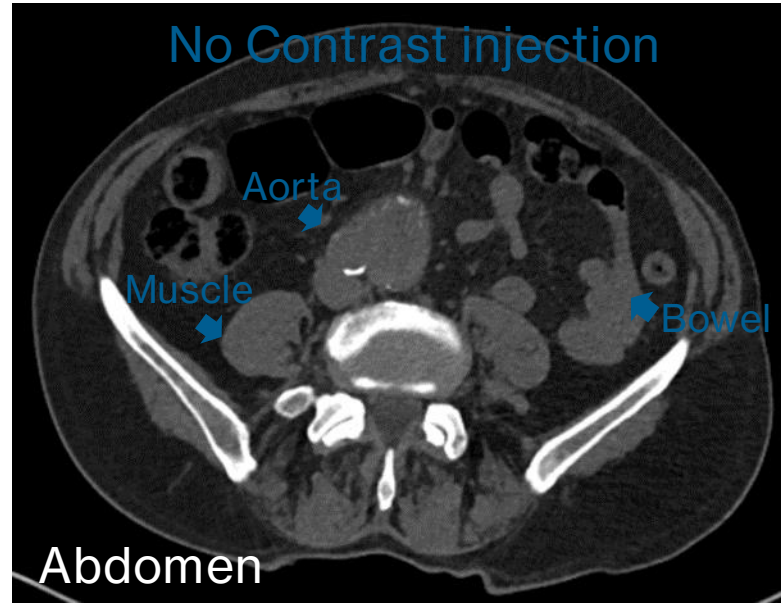
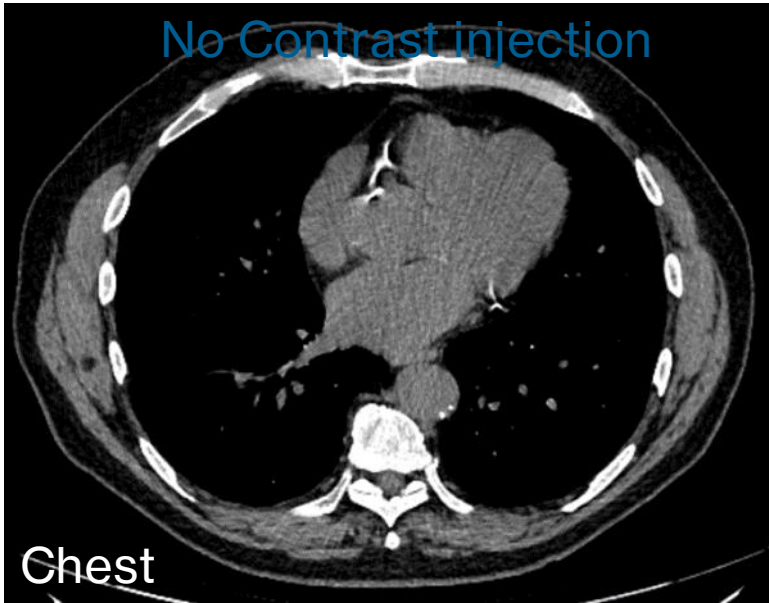
Scientific Coordinator – **NetZeroAICT** Consortium

Disclosure: R Lee is the Co-Founder and Chief Medical Officer of AiSentia

NetZeroAICT



> 300 million CT scans are performed globally / year



A non-contrast CT contains much information that human eyes can't see.

~**60% of CT scans** have intravenous iodinated radiocontrast media (**RCM**) injection to obtain **Contrast-Enhanced CT scans**

Conventional Patient Journey for a Contrast Enhanced CT

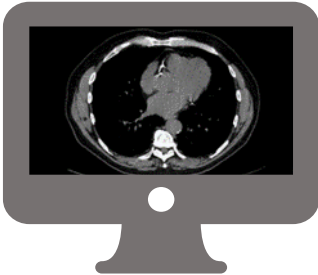
Needle
IV line



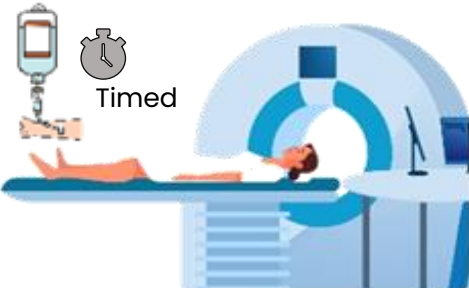
1st scan
no injection



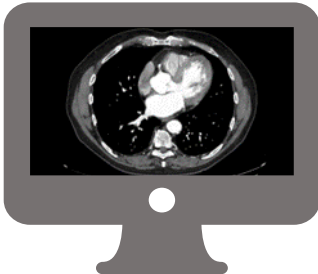
Non-Contrast
CT



2nd scan
contrast injection



Contrast
CT



Problems associated with the intravenous injection of contrast agent



Needle ('Drip') insertion-related complications:

- Pain and discomfort
- Bleeding, haematoma
- Contrast leakage out of the drip: skin necrosis
- Injury of the nearby artery: pseudoaneurysm, ischaemia

Contrast related complications:

- Contrast allergy / anaphylactoid reactions
- Kidney toxicity:
 - contrast-induced nephropathy
 - renal failure requiring dialysis/death (rare but devastating)

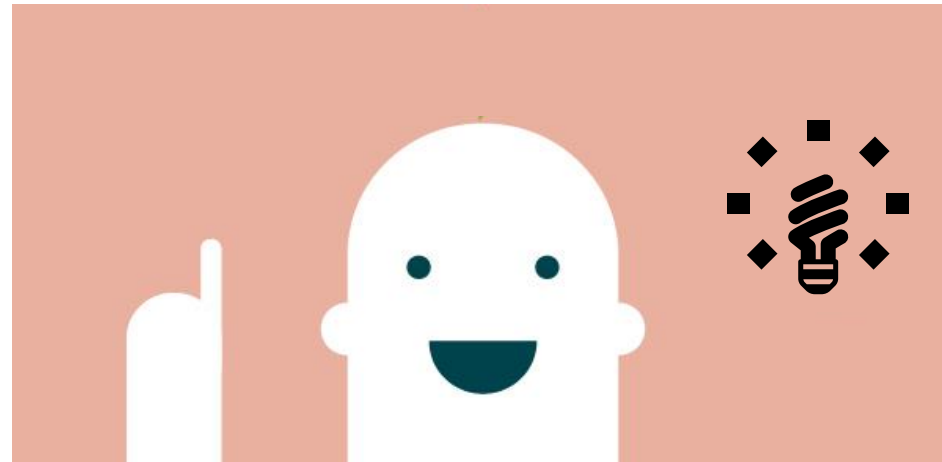
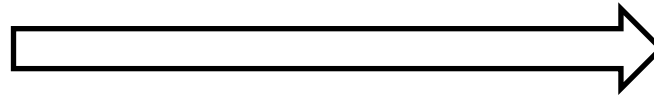
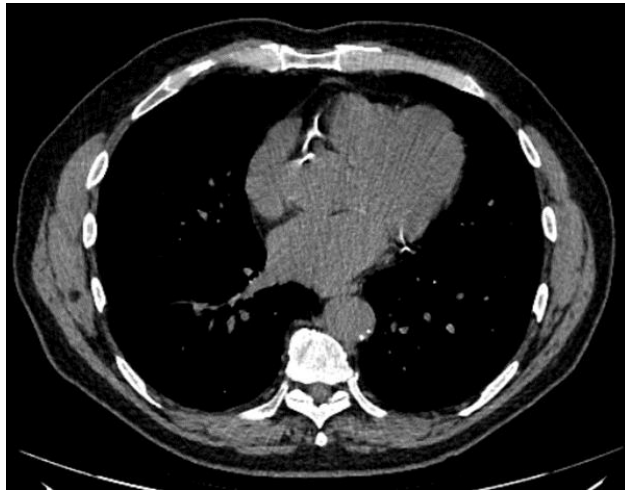


Cost of service provision:

Contrast agent, longer scan, management of complications

Lack of supply chain resilience

What if we could generate contrast enhanced CT images without the need to inject intravenous contrast?



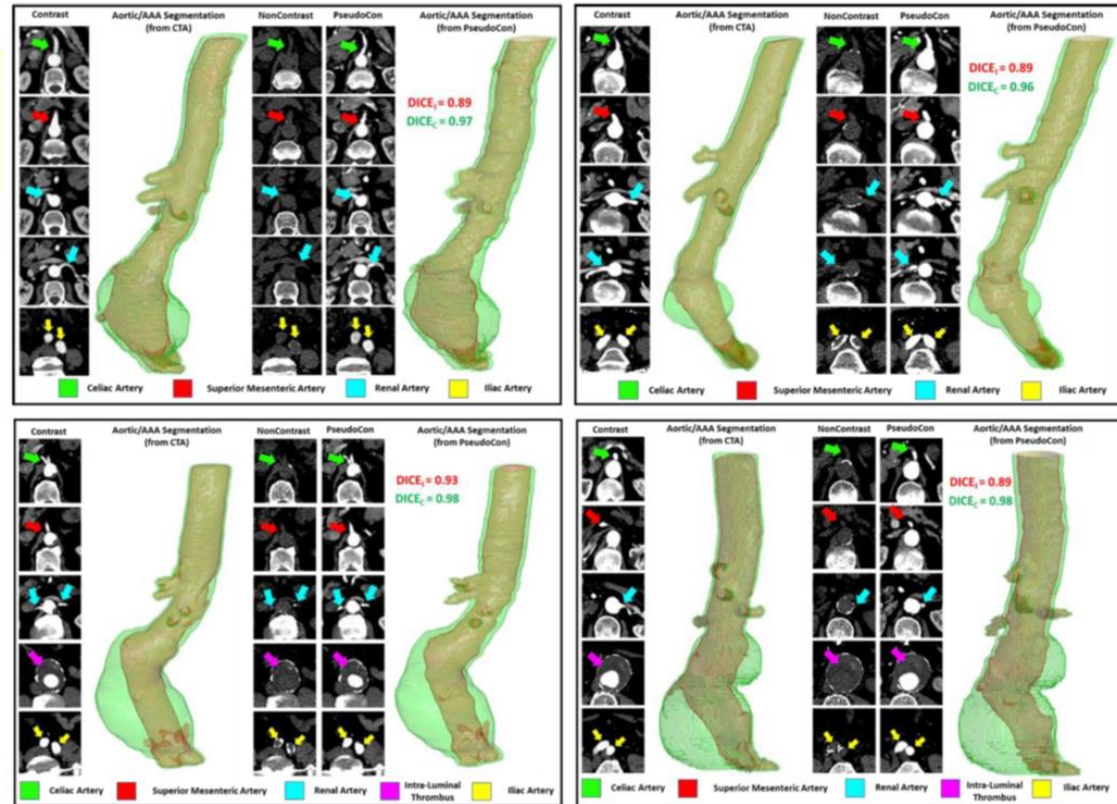
ORIGINAL ARTICLE: PDF ONLY

A Deep Learning Approach to Visualise Aortic Aneurysm Morphology without the Use of Intravenous Contrast Agents

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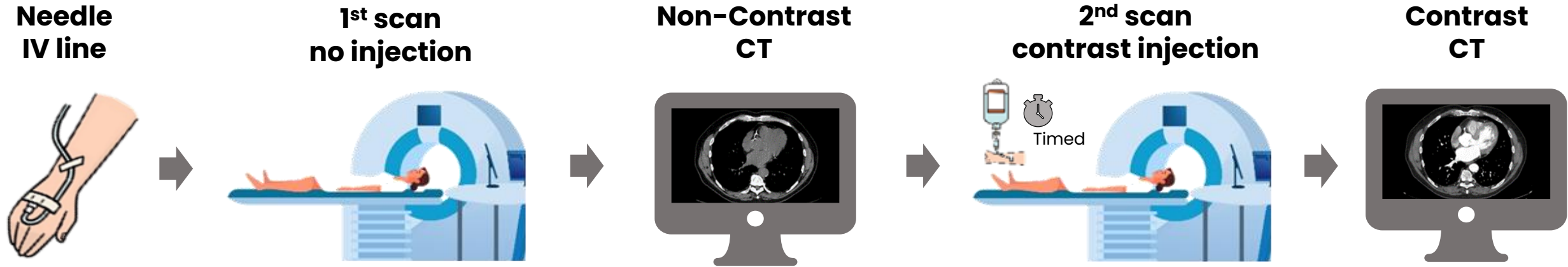
4. [WO/2021/038202](#) COMPUTERISED TOMOGRAPHY IMAGE PROCESSING

WO - 04.03.2021

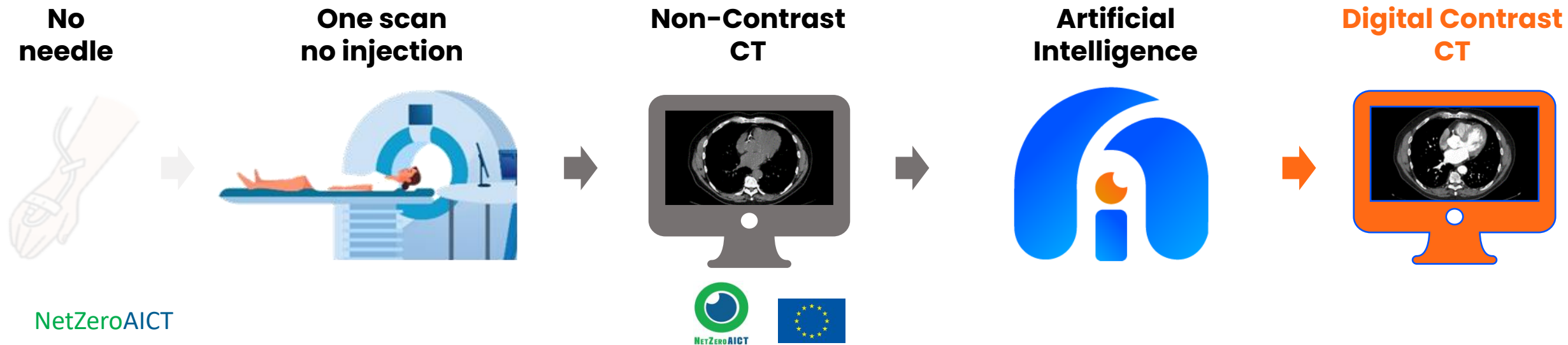
Int.Class [G06T 7/00](#)  Appl.No PCT/GB2020/052013 Applicant OXFORD UNIVERSITY INNOVATION LIMITED Inventor LEE, Regent

Methods for training an algorithm to identify structural anatomical features, for example of a blood vessel, in a non-contrast computed tomography (NCT) image are described herein. The algorithm may comprise an image segmentation algorithm, a random forest classifier, or a generative adversarial network in examples described herein. In one embodiment, a method comprises receiving a labelled training set for a machine learning image segmentation algorithm. The labelled training set comprising a plurality of NCT images, each NCT image of the plurality of NCT images showing a targeted region of a subject, the targeted region including at least one blood vessel. The labelled training set further comprises a corresponding plurality of segmentation masks, each segmentation mask labelling at least one structural feature of a blood vessel in a corresponding NCT image of the plurality of NCT images. The method further comprises training a machine learning image segmentation algorithm, using the plurality of NCT images and the corresponding plurality of segmentation masks, to learn features of the NCT images that correspond to structural features of the blood vessels labelled in the segmentation masks, and output a trained image segmentation model. The method further comprises outputting the trained image segmentation model usable for identifying structural features of a blood vessel in an NCT image. Further methods are described herein for identifying anatomical features from an NCT image, and for establishing training sets. Computing apparatuses and computer readable media are also described herein.

Conventional Patient Journey for a Contrast Enhanced CT



Patient Journey for a 'CT Digital Contrast'



Curbing the carbon footprint of health care

Technology and innovation is a key theme of the 2023 UN Climate Change Conference, COP28, hoping to catalyse the development and deployment of climate solutions to drive down global temperatures and protect human health. The health-care sector contributes approximately 4.6% to global greenhouse gas (GHG) emissions. Will wider adoption of digital health technologies help mitigate this issue, or exacerbate it?

Around 1% of global GHG emissions is attributable to medical imaging, which is both resource-intensive and energy-intensive. Artificial intelligence (AI) applications in medical imaging have been transformative, from assisting breast cancer screening to computer-aided detection for tuberculosis, but there are concerns that AI

detected during colonoscopy should be resected and, if so, whether they should be retrieved for histopathological examination. Reducing the number of unnecessary resections and biopsies can further reduce the health-care sector's GHG emissions.

Reducing the number of in-person consultations is another effective means of lowering health-care emissions. Telemedicine has become a vital tool in the path towards more sustainable health care by reducing the need for patient and provider travel. Availability of virtual consultations has been shown to reduce GHG emissions of health-care systems, and is also associated with lower physician burnout, which might consequently improve the quality of patient care.



alengo/Getty Images

For more on COP28 see <https://www.cop28.com/en/>

For the 2023 report of the **Lancet Countdown** see **Countdown** 2023; published online Nov 14. [https://doi.org/10.1016/S0140-6736\(23\)01859-7](https://doi.org/10.1016/S0140-6736(23)01859-7)

For more on the **carbon costs of medical imaging** see *J Clin Med* 2023; 12: 215

For more on the **carbon**

Contrast CT and environmental impact

Over **180 million** contrast CT scans performed globally per year generating combined sizable carbon footprints



Single use packaging

42,000
tons



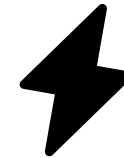
Needle (steel)

900
tons



Plastic waste

90,000
tons



Energy consumption

150,000,000
kWh



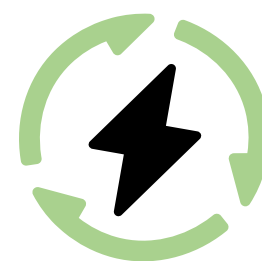
Iodine in wastewater

200,000
tons

Average CO₂ emission per CT scan is 9.2kg*!

Digital Contrast – Towards Sustainable and Climate Neutral Health Systems

With a target of reducing 30% of contrast CT scans



Digital Contrast can reduce CO₂ emission, per year

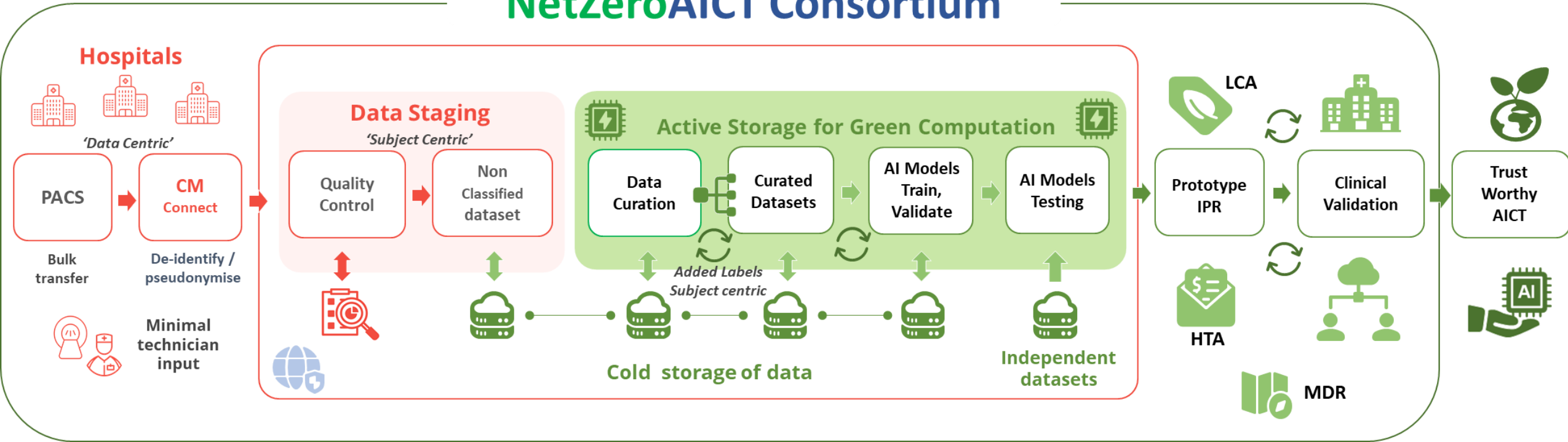
EU 300,000 tons

Global 6–900,000 tons

(↓ global iodine wastewater contamination by **70,000 tons**)



NetZeroAICT Consortium



NetZeroAICT



Legal Framework / GDPR



Ethics / IRB approval



Stakeholder involvement

Trustworthiness as a core mission

NetZeroAICT



The most diverse dataset in a centralised CT Scan repository

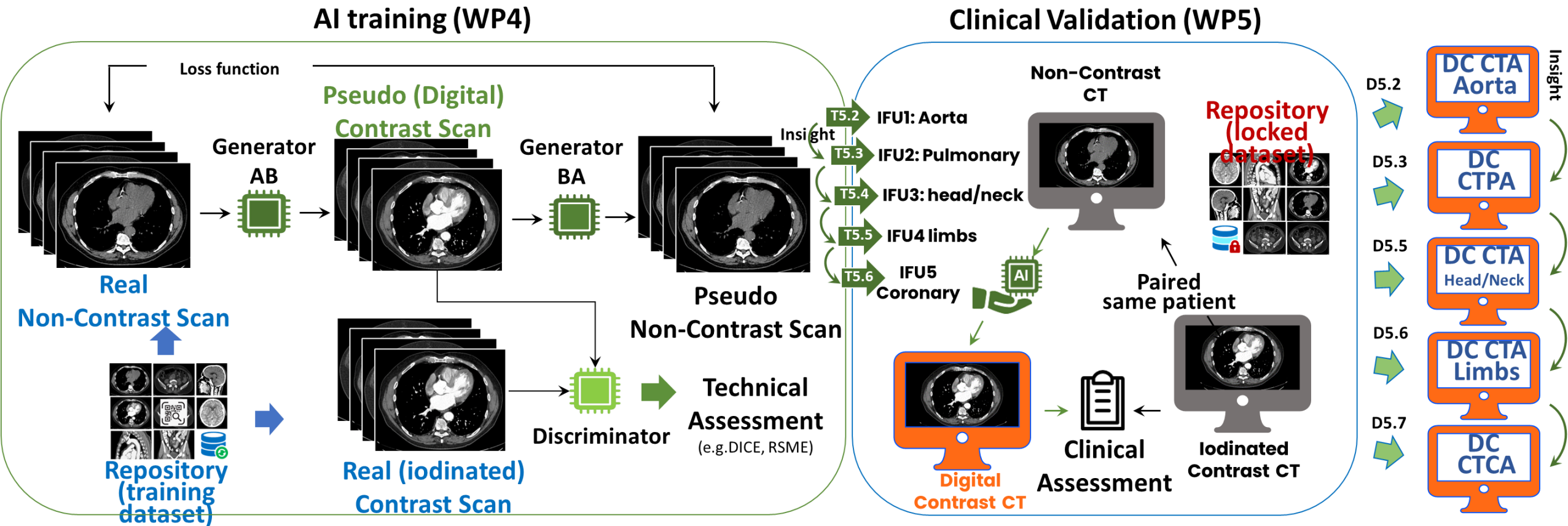
(>1 million scans, 3 continents)



Pseudonymised scans
'unsorted'

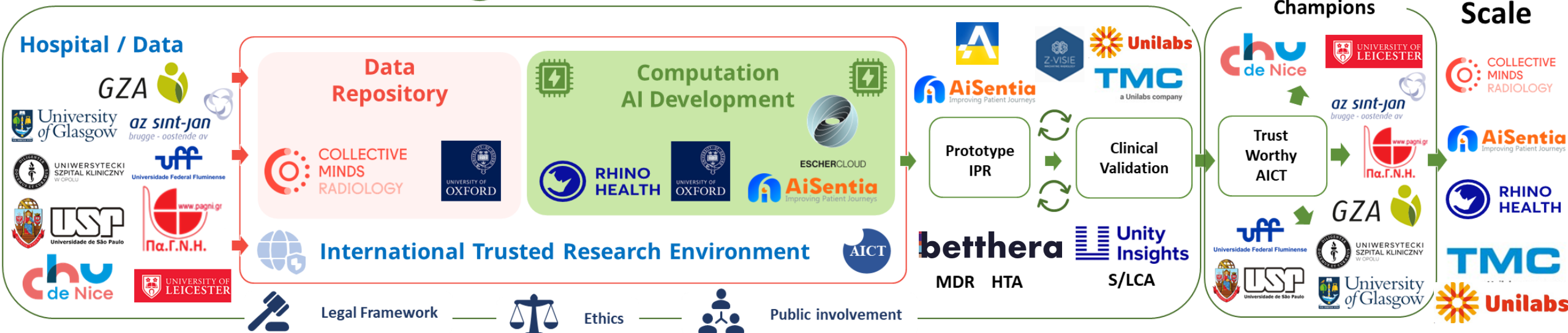
curated, catalogued, 'research ready'

Train and validate CT angiogram applications





NetZeroAICT Value Chain



NetZeroAICT Consortium



