

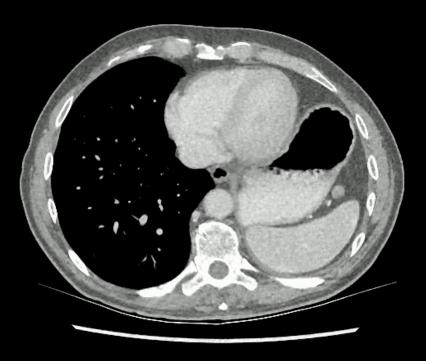
# Scaling Annotations, Datasets, and Algorithms for Medical Image Analysis

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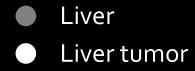
## Radiologists hate annotation, but computer scientists love it.

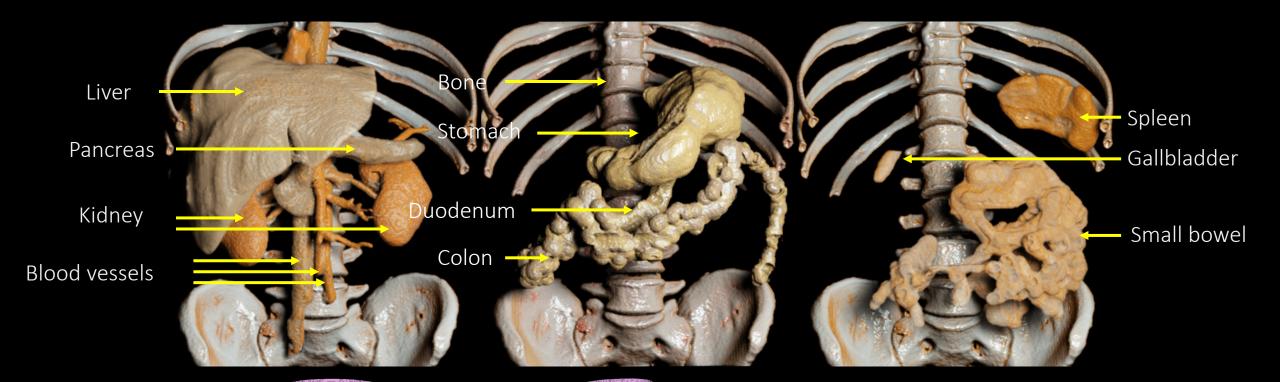
CT



Ground Truth annotated by human experts

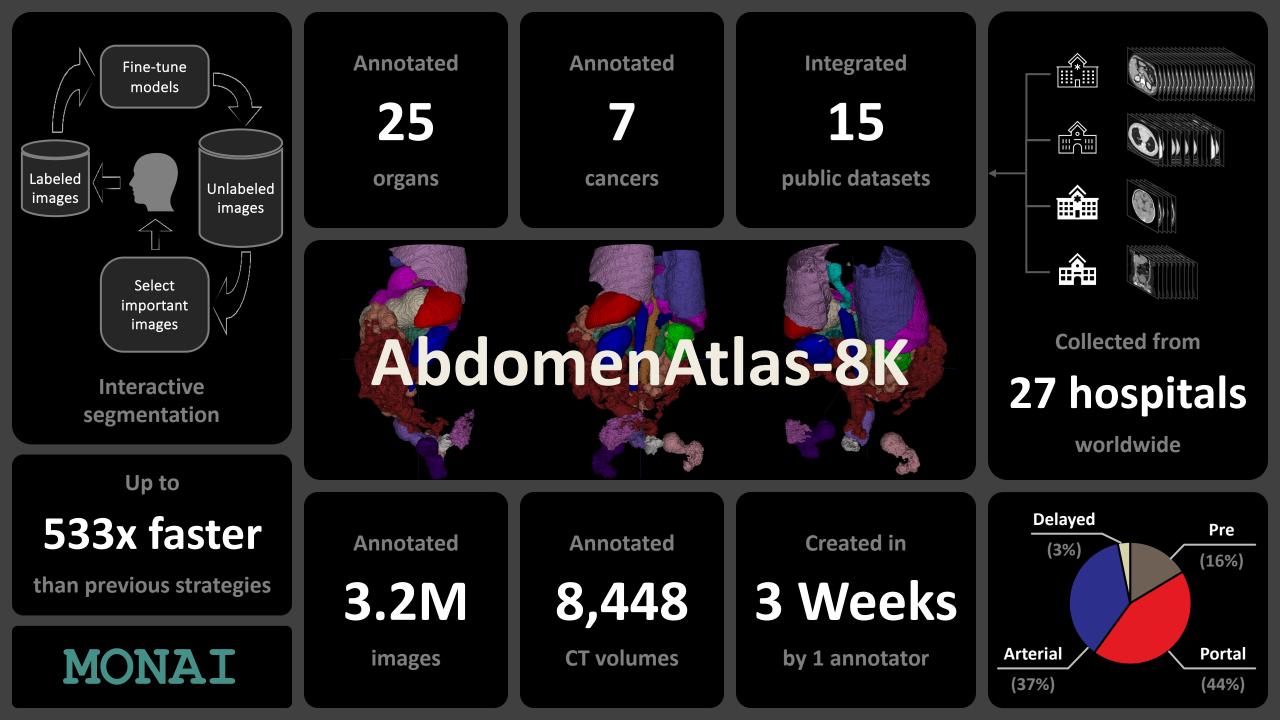
Deep Learning

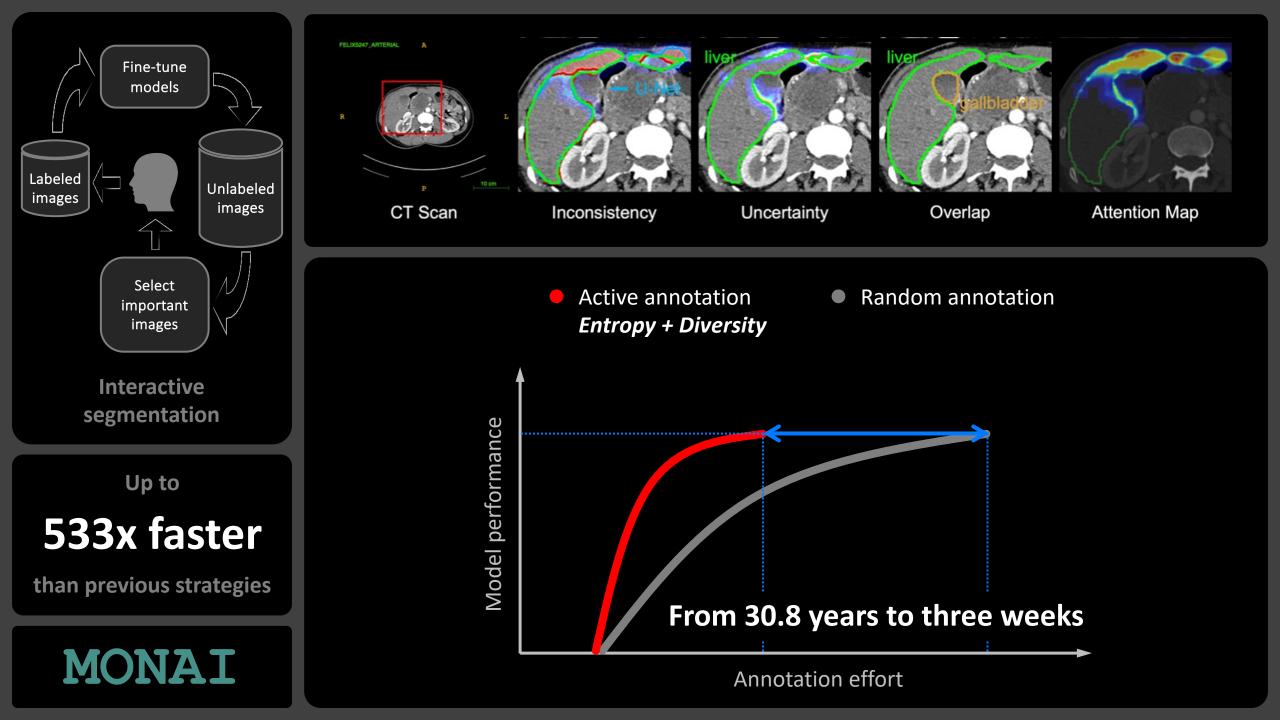


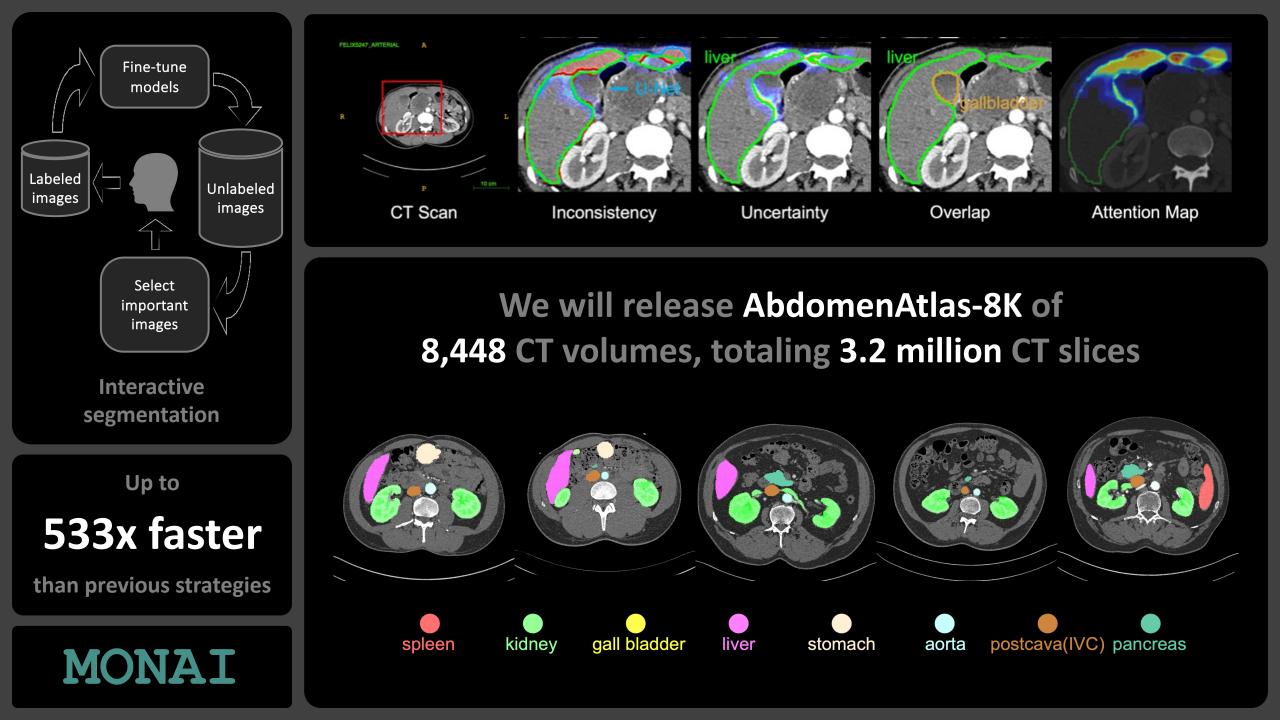


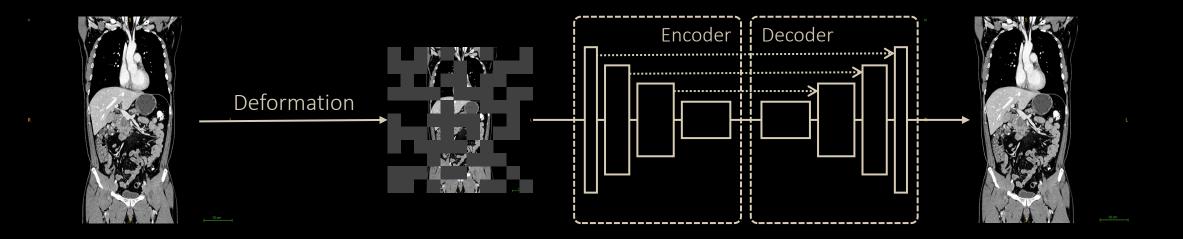
# Body Maps: 3D Maps of Whole Body

# Scaling <u>Annotations</u>, Datasets, and Algorithms for Medical Image Analysis

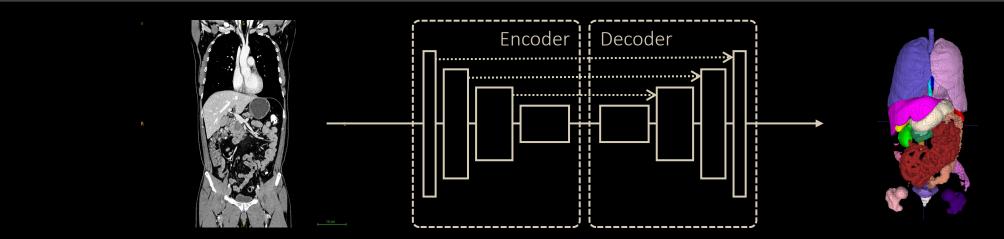








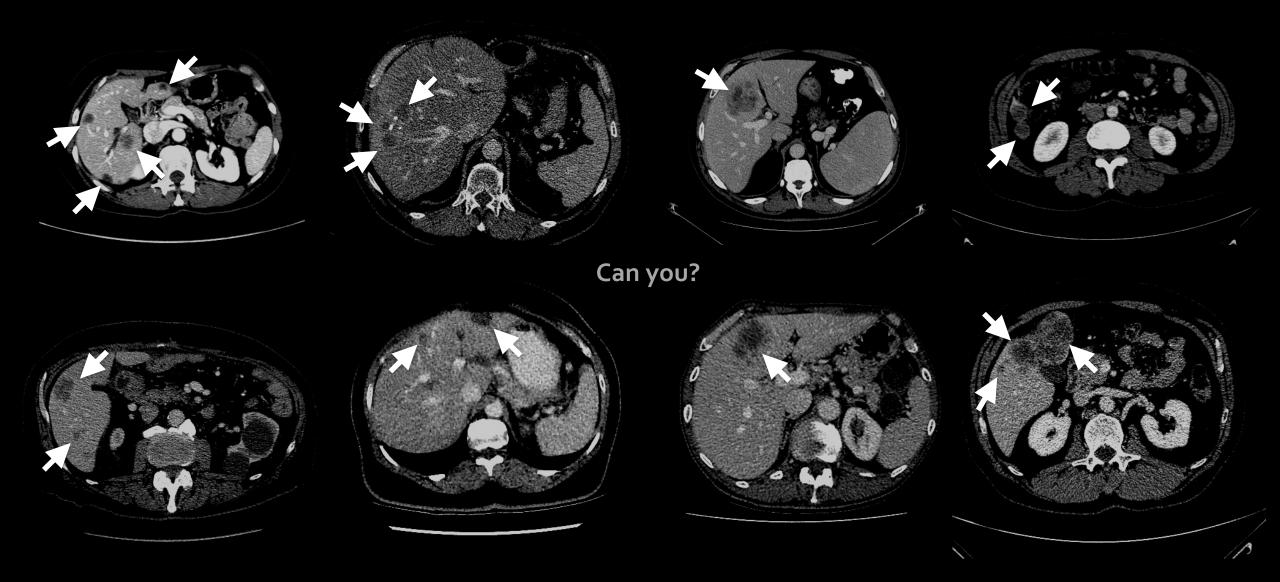
# Self-supervised Pre-training 5,000 data + 1,152 GPU hours



Supervised Pre-training 20 data + 20 annotation + 40 GPU hours

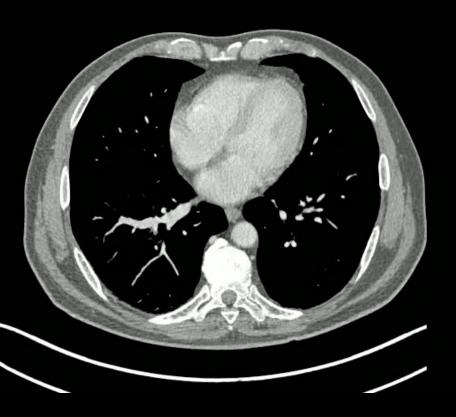
# Scaling Annotations, <u>Datasets</u>, and Algorithms for Medical Image Analysis

## Medical professionals cannot tell which are real and which are synthetic tumors

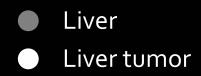


### Training AI on synthetic tumors performs as well as training it on real tumors

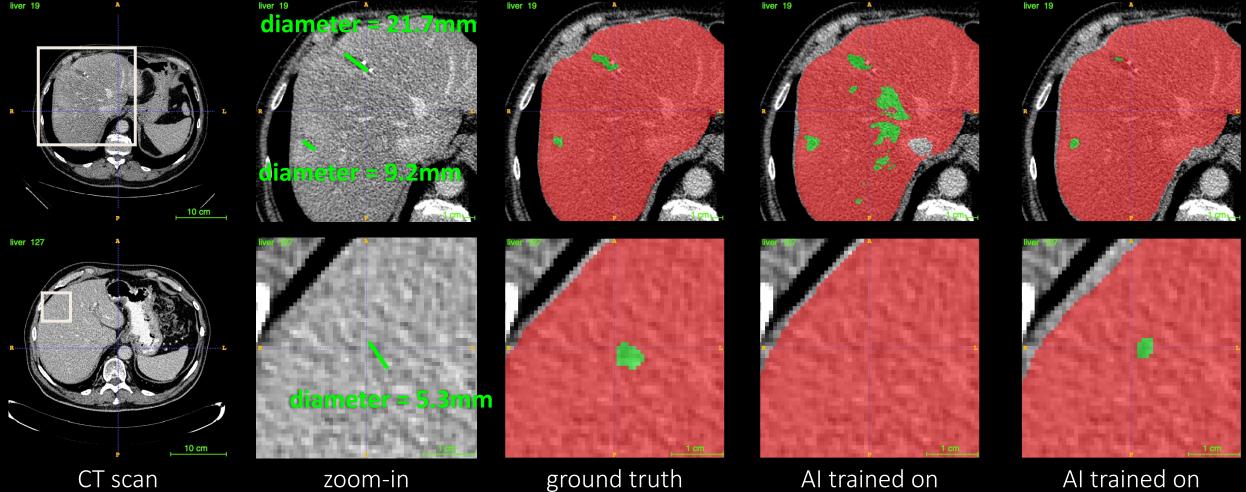
CT



Al prediction trained on real tumors with per-voxel annotation DSC = 58% [52% - 63%] Al prediction trained on synthetic tumors *with no annotation* DSC = 60% [55% - 65%]



## [Qualitative] Generating enormous small tumors for training AI models

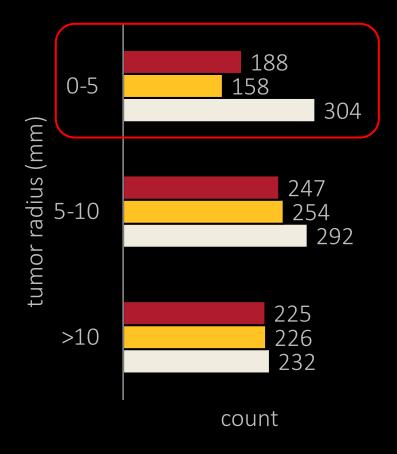


Al trained on synthetic tumors

real tumors

# [Quantitative] Generating enormous small tumors for training AI models

Al trained on synthetic tumorsAl trained on real tumorsground truth



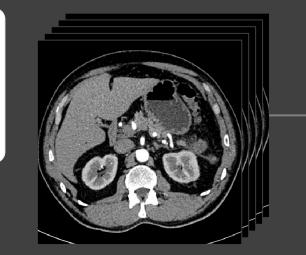
Observation: Compared with real tumors, Al trained on synthetic tumors improves Sensitivity from 52% to 62% for detecting small tumors (0-5mm).

- Needed for early detection
  - Early signs of cancer can be subtle
  - o 1/2 of liver cancer are missed by radiologists
- Needed for AI development
  - CT scans with early cancer are limited
  - Annotations for early cancer are hard
- Needed for medical education
  - Junior radiologists have an Accuracy of 20%
  - Senior radiologists have an Accuracy of 78%

# Scaling Annotations, Datasets, and <u>Algorithms</u> for Medical Image Analysis

featured in ChimeraX

at UCSF



Vision Encoder

Please segment the tumor in the tail of the pancreas and then measure its size.

Take a look at these CT scans and mark the suspected tumor region.

Universal Model

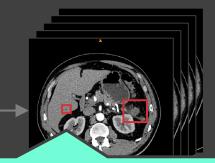
Text Encoder

featured in MONAI

at NVIDIA

This tumor is likely to be PDAC with a diameter of 25mm.



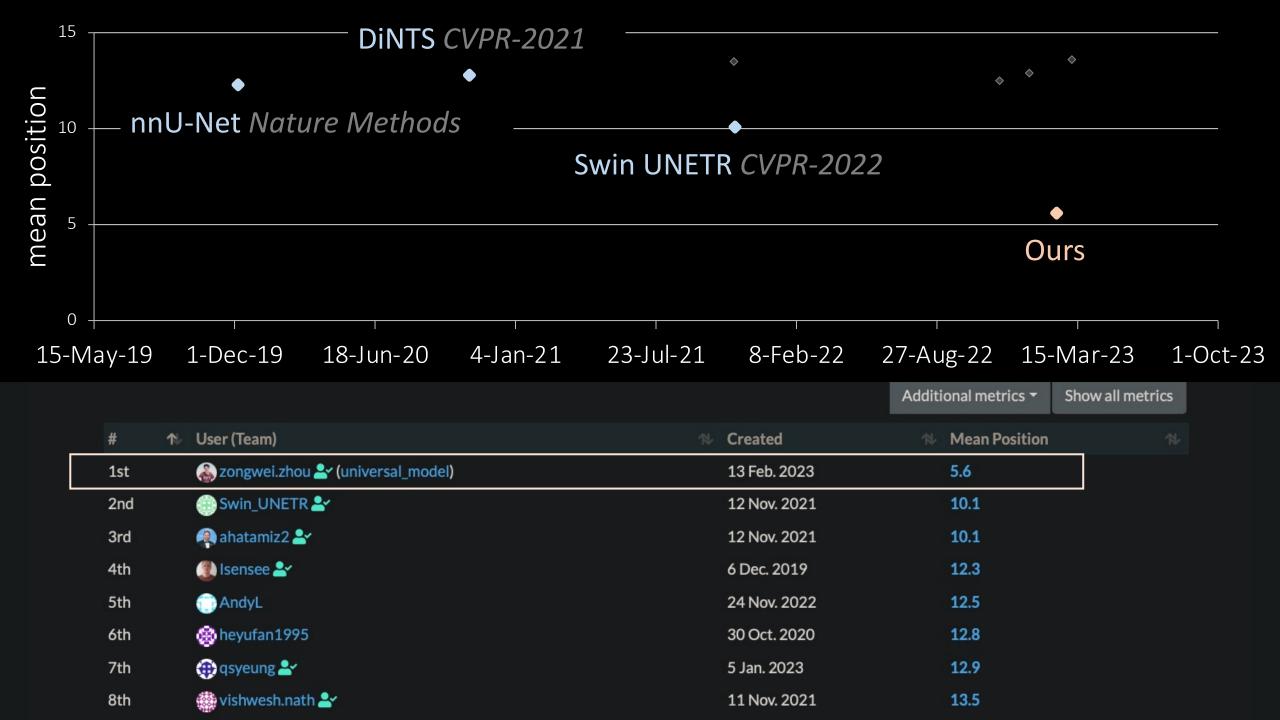


Two potential tumors are framed in bounding boxes.

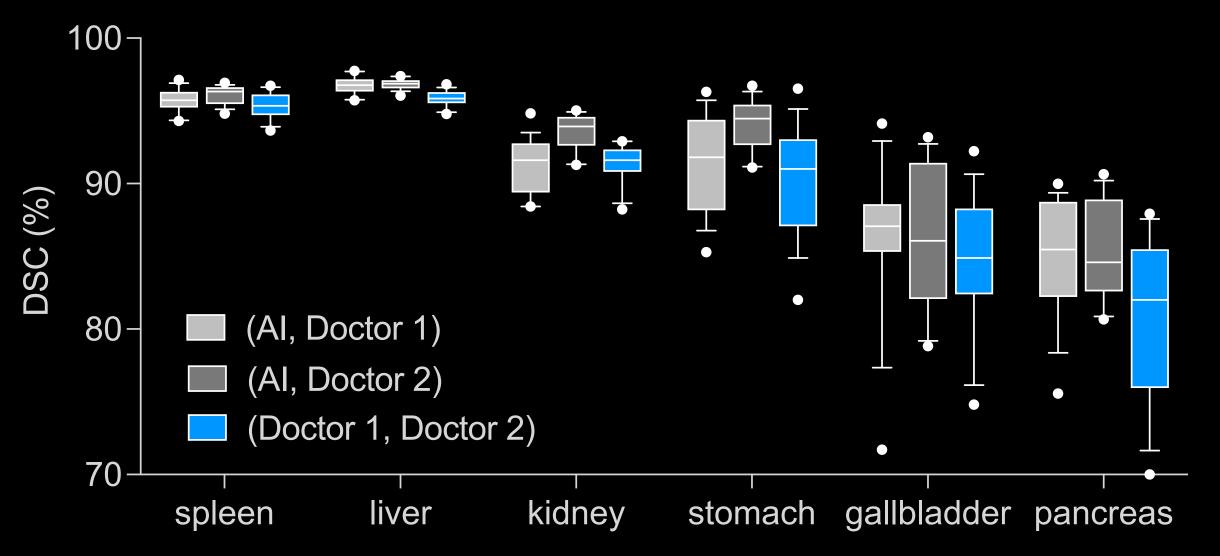
 $\overline{\mathcal{A}}$ 

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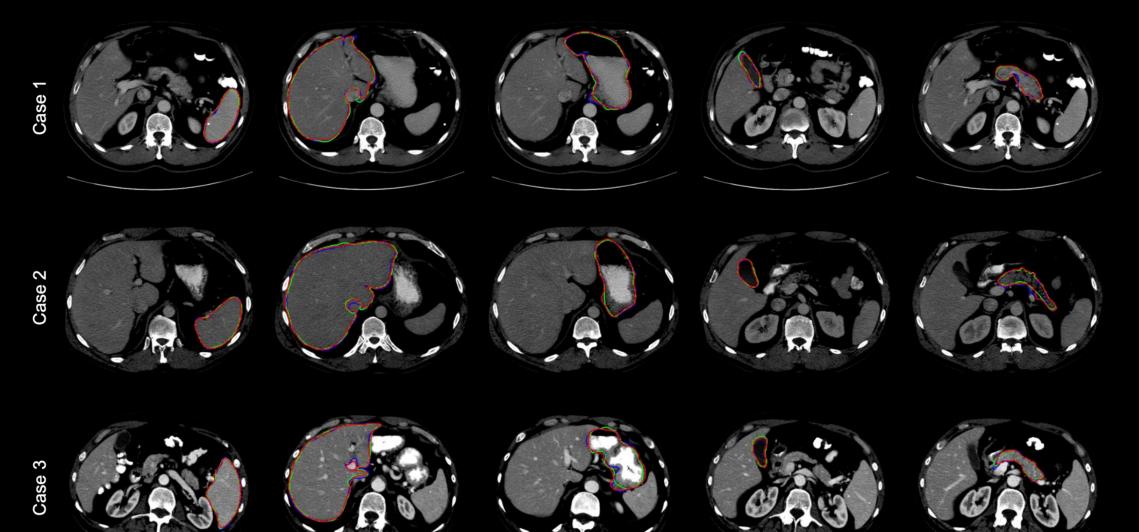
1st	嵡 zongwei.zhou 峇 (universal_model)	13 Feb. 2023	5.6
2nd	🌐 Swin_UNETR 峇	12 Nov. 2021	10.1
3rd	👰 ahatamiz2 峇	12 Nov. 2021	10.1
4th	🕘 Isensee 峇	6 Dec. 2019	12.3
5th	AndyL	24 Nov. 2022	12.5
6th	🛞 heyufan 1995	30 Oct. 2020	12.8
7th	💮 qsyeung 🛓	5 Jan. 2023	12.9
8th	🎆 vishwesh.nath 峇	11 Nov. 2021	13.5



The AI predictions for six organs are comparable to expert annotators If we spend a lot more money to ask radiologists to annotate these six organs, it might turn out that the AI can do a similar quality annotation



The AI predictions for six organs are comparable to expert annotators If we spend a lot more money to ask radiologists to annotate these six organs, it might turn out that the AI can do a similar quality annotation



#### **Scaling annotations**

Efficient annotation Human in the loop Novel disease

# AbdomenAtlas-8K

8,448 annotated CT volumes

#### **Scaling datasets**

Multiple modalities Diverse institutes IRB approval

# Tumor <mark>Syn</mark>thesis

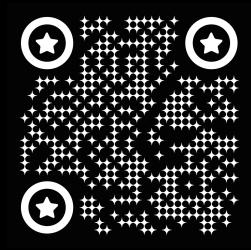
Annotation-free deep learning

#### **Scaling algorithms**

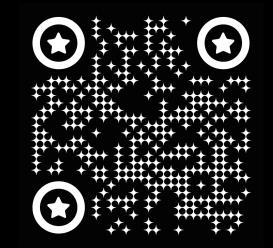
Vision-language Lifelong learning Reader study

# **Universal Model**

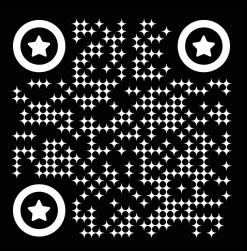
25 organs and 7 cancers



Code & Dataset NeurIPS-2023



Code & Turing Test CVPR-2023



Code & Model ICCV-2023

#### Reference

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