

Motivation

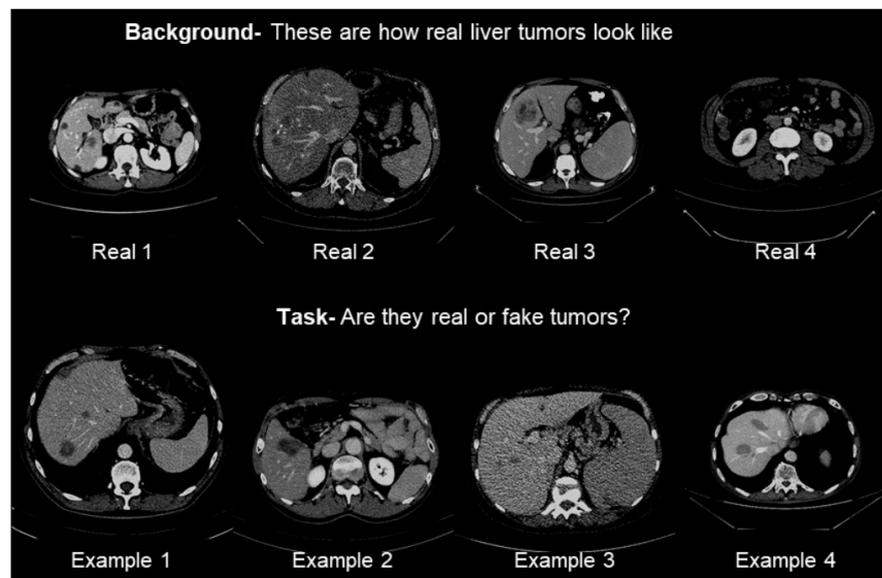
Problem

Artificial intelligence (AI) has dominated medical image segmentation, but training an AI model often requires a large number of **detailed per-voxel annotations**. Annotating medical images is not only expensive and time-consuming, but also requires **extensive medical expertise**.

Solution

To reduce annotation costs, generating **synthetic tumors** is an emerging research topic. In this paper, we develop a novel strategy to generate synthetic liver tumors. Our strategy have two **intriguing** advantages:

- Realistic in shape and texture, which even **medical professionals** can confuse with real tumors;
- Effective for training AI models, which can perform liver tumor segmentation **similarly** to the model trained on real tumors



Visual Turing Test

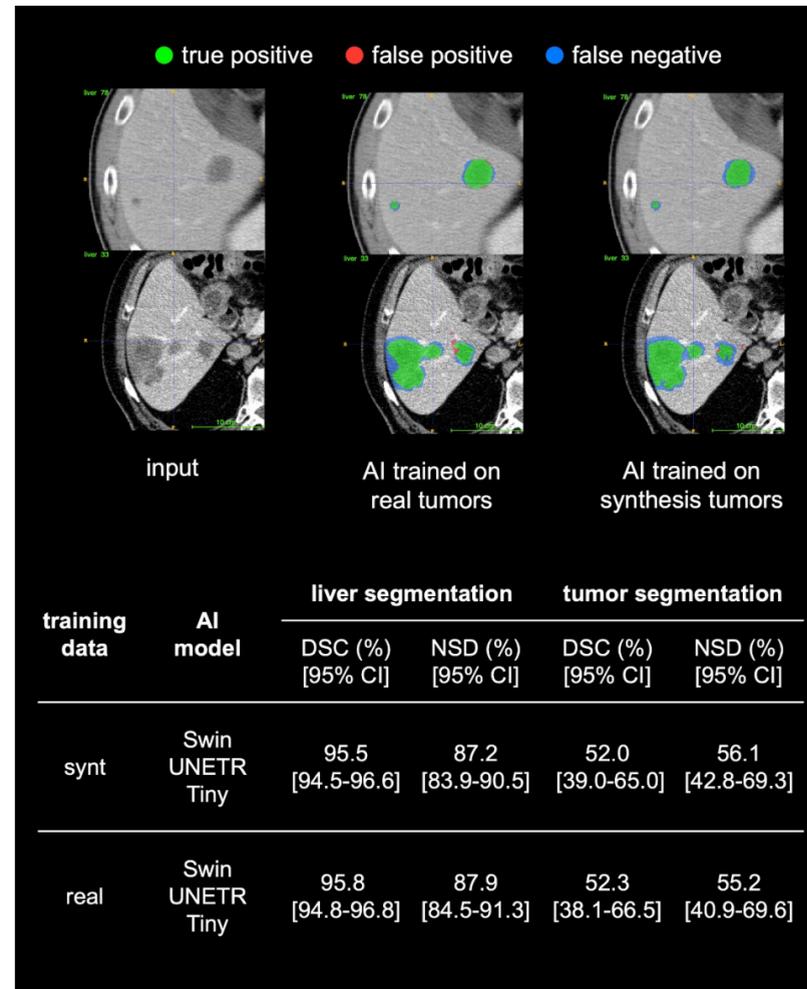
Experiment Setup

Task: Liver Tumor Segmentation

AI Model: Swin-UNETR-Tiny¹

Real Data: Detailed per-voxel annotations of liver tumors in the LiTS² dataset.

Synthetic Data: Healthy liver assembled from 4 different dataset.

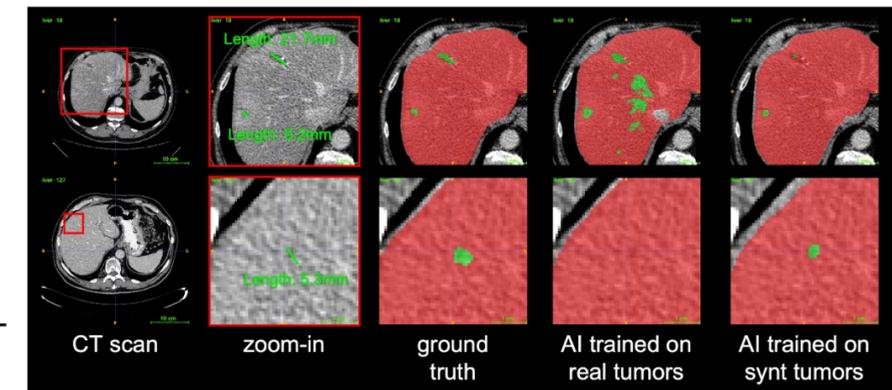
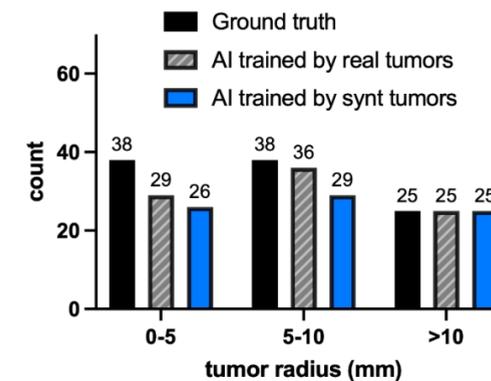


Segmentation Results

Results

- A professional with 6-year experience results show an accuracy of 60% in Visual Turing Test.
- Synthetic tumor achieves **similar** DSC and **higher** NSD compare to real tumor in tumor segmentation.
- Our synthetic tumor have the potential to facilitate small tumor detection.

This result is **unprecedented** because no existing work, using synthetic tumors only, has thus far reached a similar or even close performance to the model trained on real tumors.



Small Tumor Detection

Reference:

- [1] Tang, Yucheng, et al. "Self-supervised pre-training of swin transformers for 3d medical image analysis." *Proceedings of the IEEE/CVF Conference on Computer Vision and Pattern Recognition*. 2022.
- [2] Bilic, Patrick, et al. "The liver tumor segmentation benchmark (lits)." *Medical Image Analysis* (2022): 102680.