

Annotation-Efficient Deep Learning for Medical Image Analysis

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Statistics

6,458 citations

Top 2% of Scientists in 2022

24 first/corresponding authored papers



Significant

Applications
Methodologies

Impactful



2017-2021
Arizona State University
PhD Degree: Biomedical Informatics



Significant

Applications

Impactful

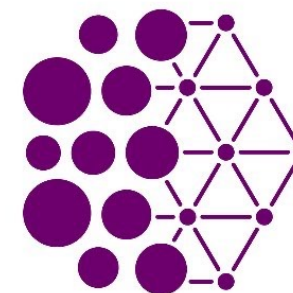
Methodologies

Bachelor's Degree: Computer Science
Dalian University of Technology
2012-2016



Postdoc: Computer Science
Johns Hopkins University
2021-present

Internship: Mayo Clinic, Mila

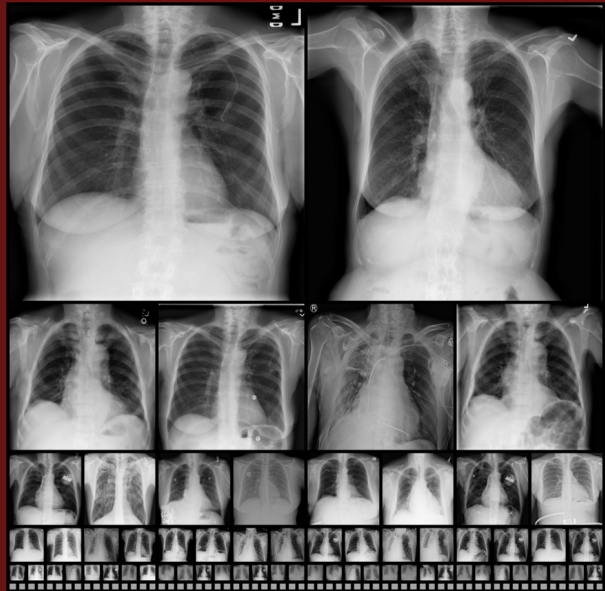


Mila

Significant

Applications
Methodologies

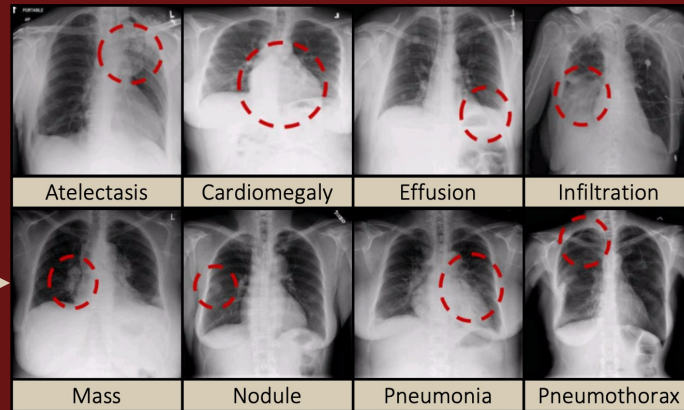
Impactful



Data



Annotate



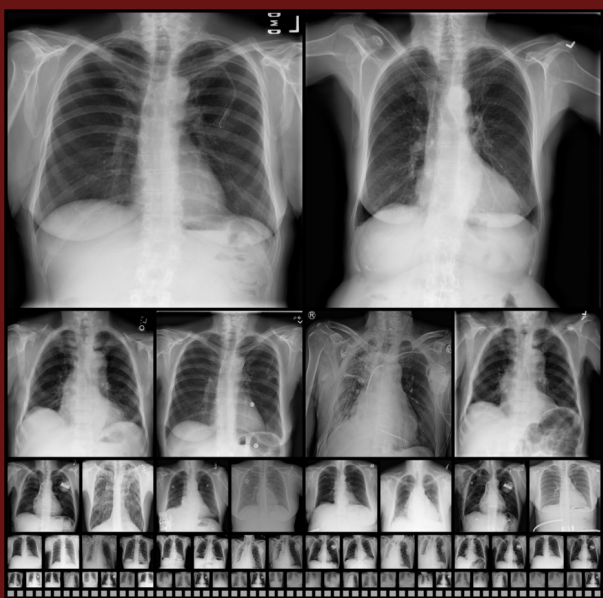
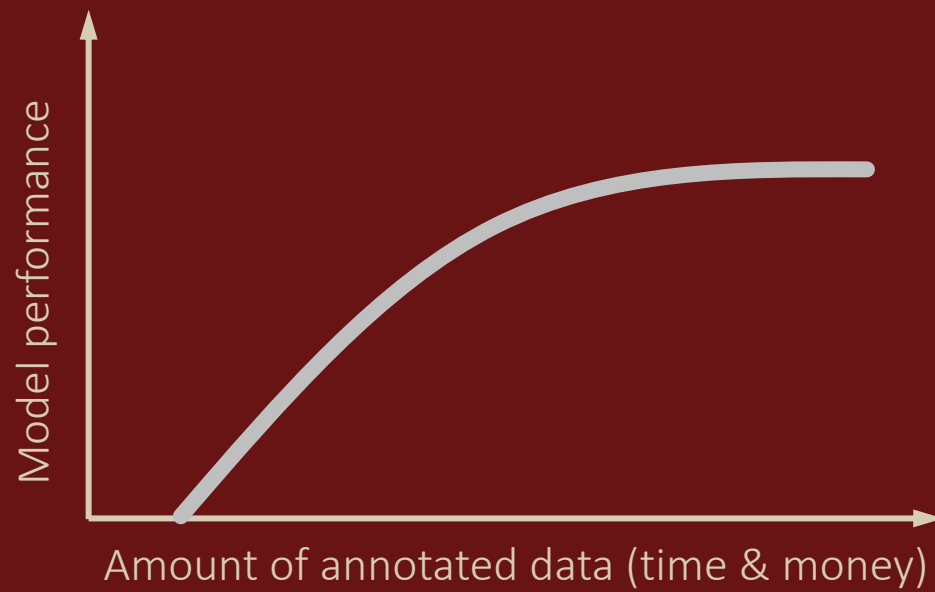
Data & Annotation



Model



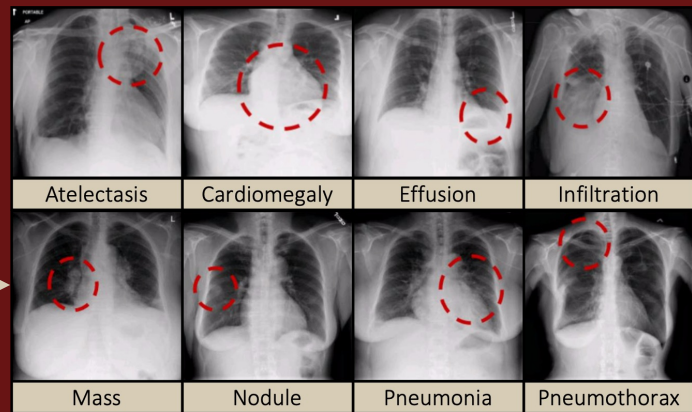
Applications



Data



Annotate



Data & Annotation

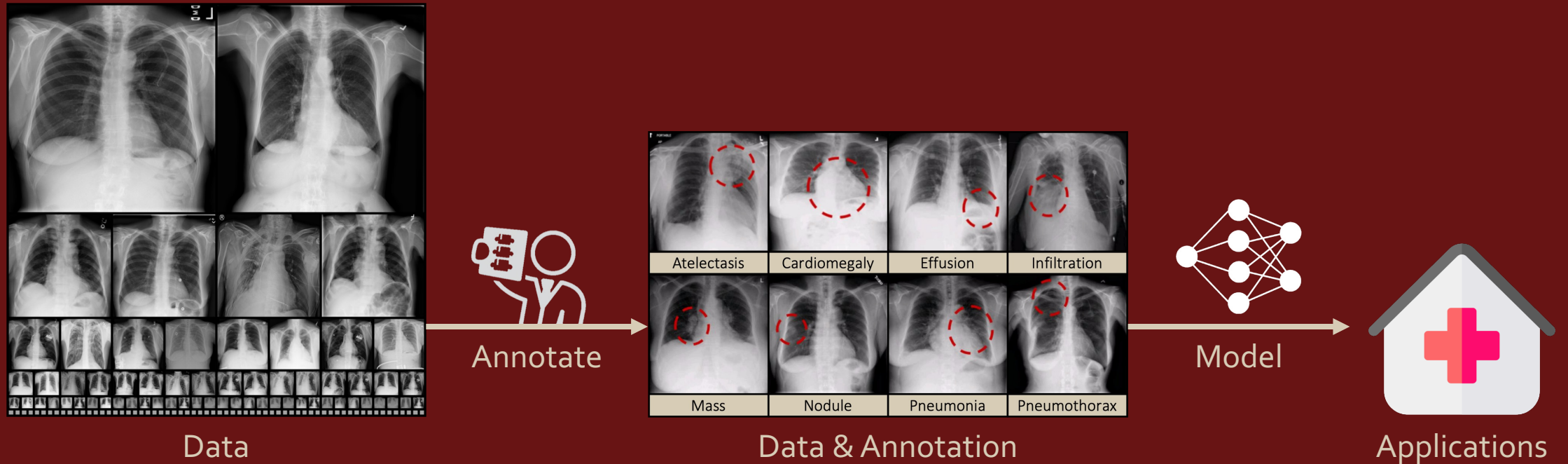
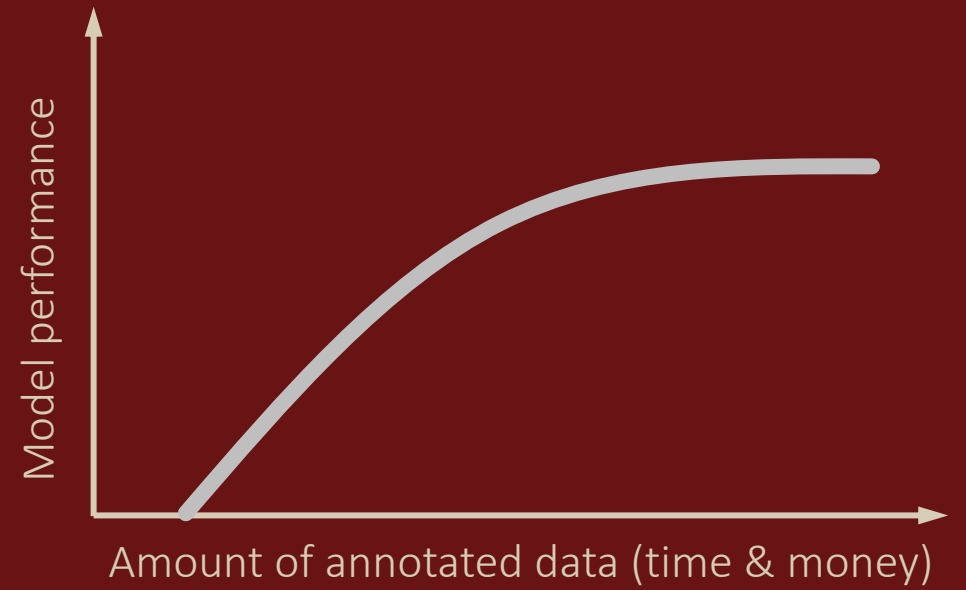


Model



Applications

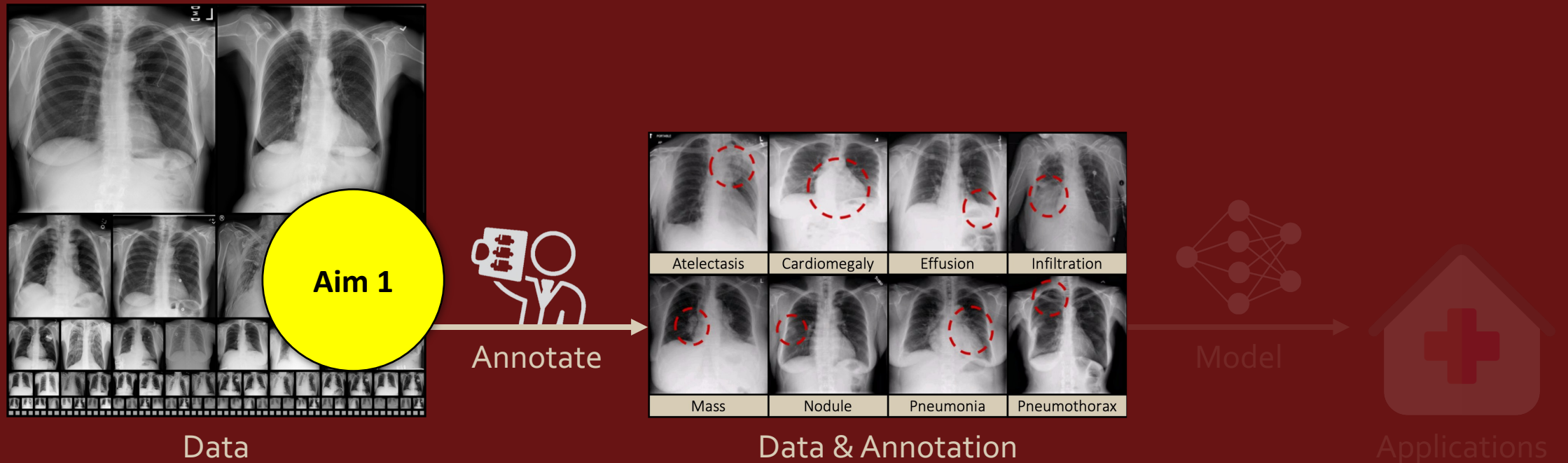
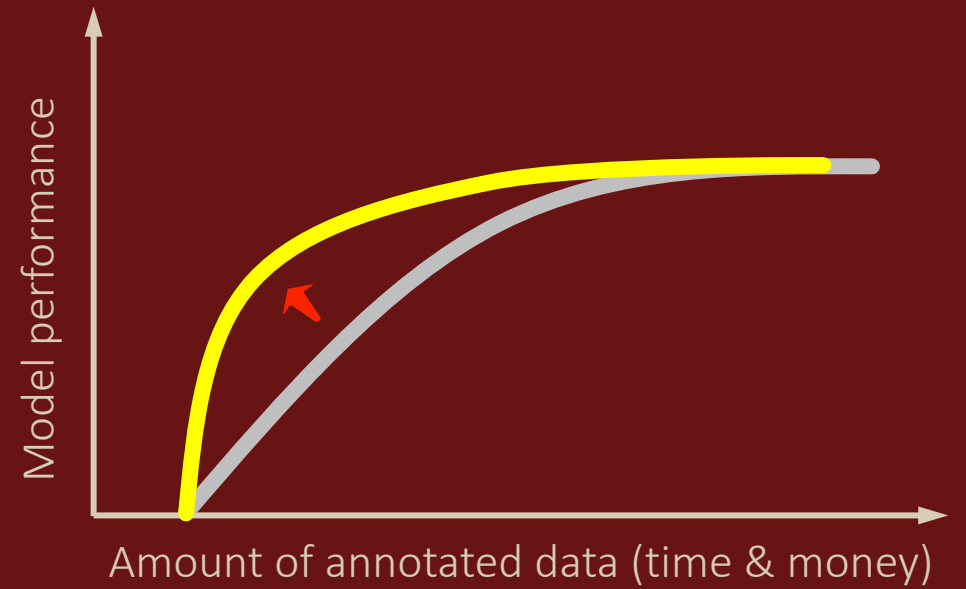
Goal: Minimize manual annotation efforts for rapid, precise computer-aided diagnosis systems.



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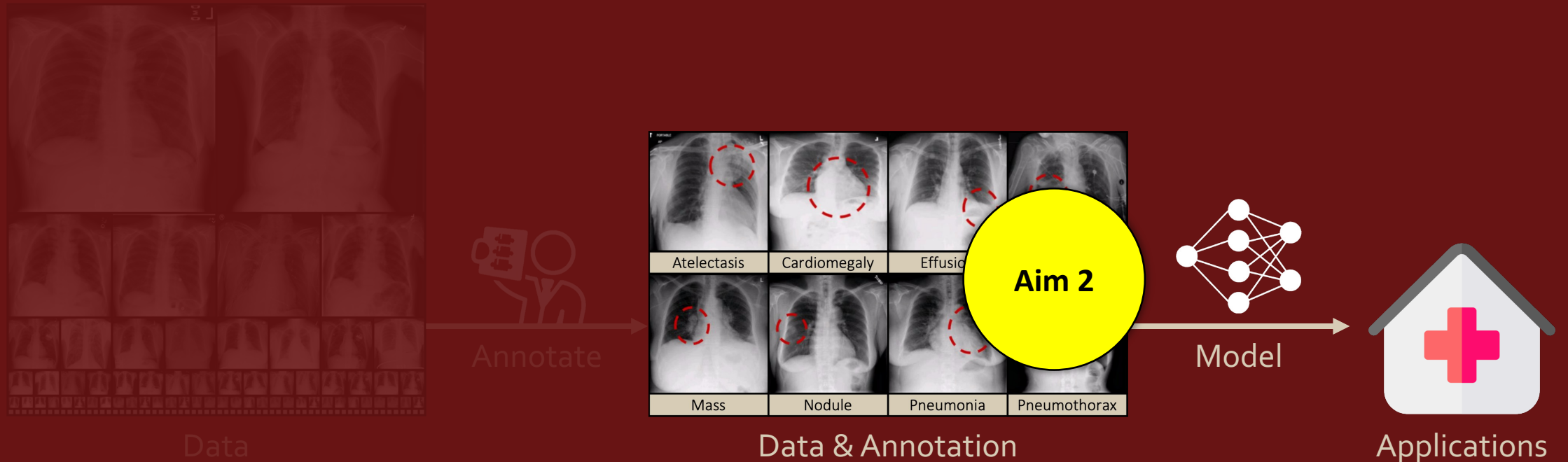
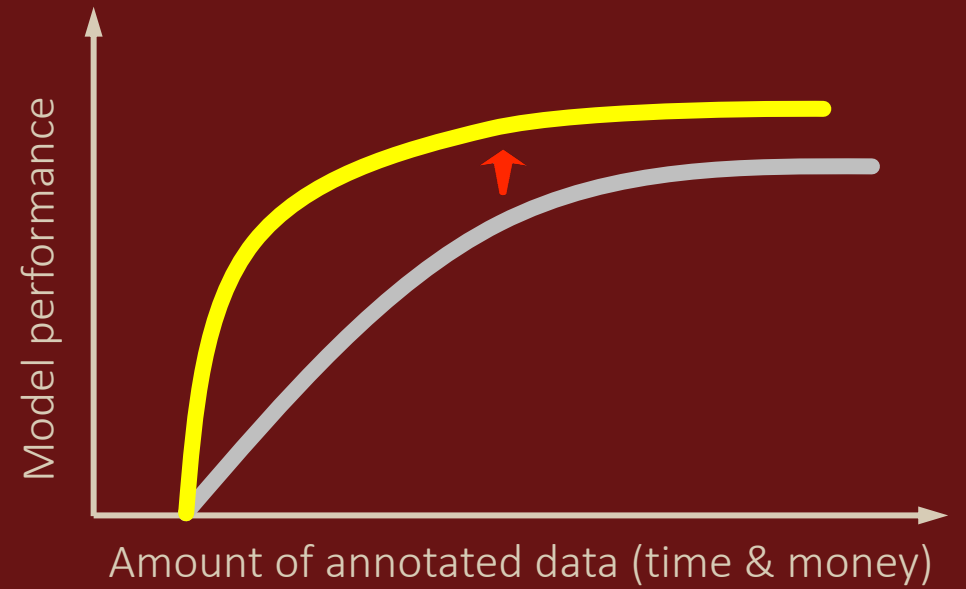
1. Acquiring necessary annotation efficiently from human experts.

- Active, Continual Fine-Tuning (ACFT)
- CVPR-2017, MedIA-2021, MIDL-2023



Goal: Minimize manual annotation efforts for rapid, precise computer-aided diagnosis systems.

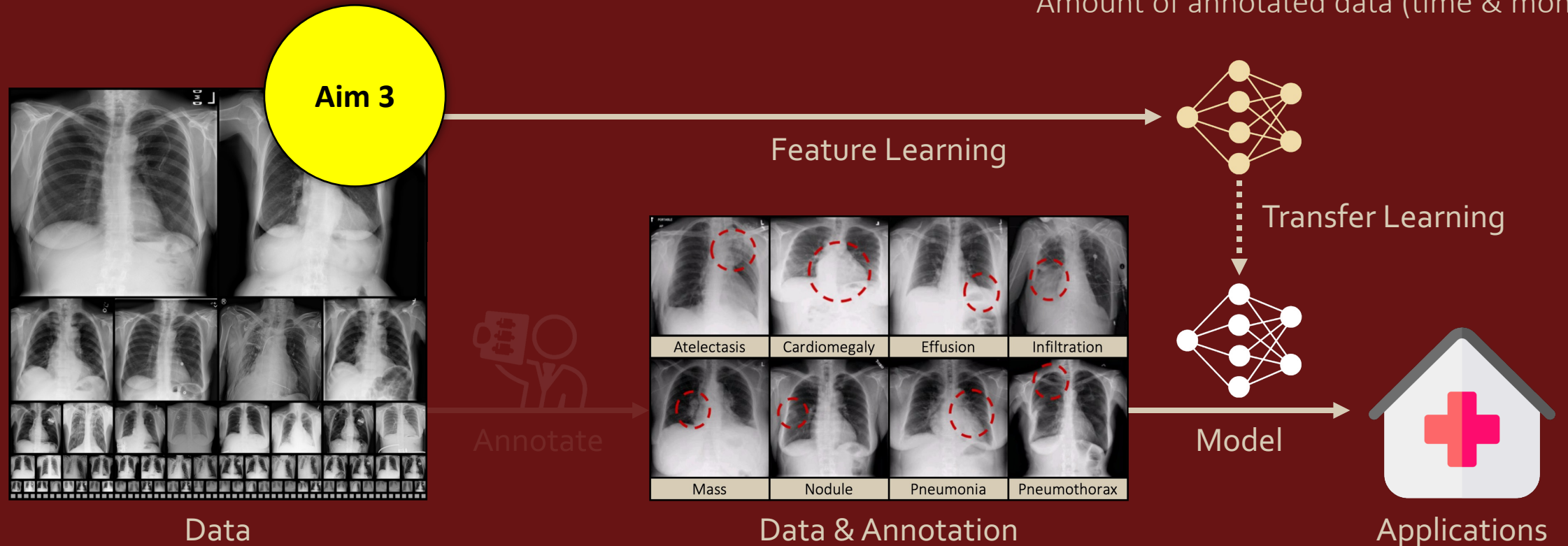
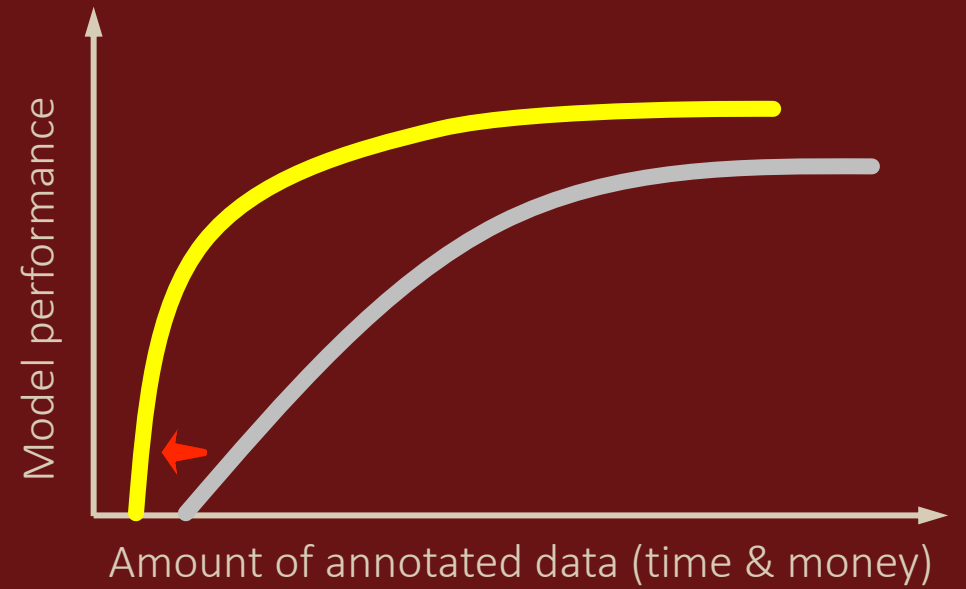
1. Acquiring necessary annotation efficiently from human experts.
2. Utilizing existing annotation effectively from advanced models.
 - UNet++
 - MICCAIW-2018, IEEE TMI-2019 (*Most Popular Articles*)



Goal: Minimize manual annotation efforts for rapid, precise computer-aided diagnosis systems.

1. Acquiring necessary annotation efficiently from human experts.
2. Utilizing existing annotation effectively from advanced models.
3. Extracting generic knowledge directly from unannotated images.

- Models Genesis
- MICCAI-2019 (*Young Scientist Award*), MedIA-2020 (*Best Paper Award*)

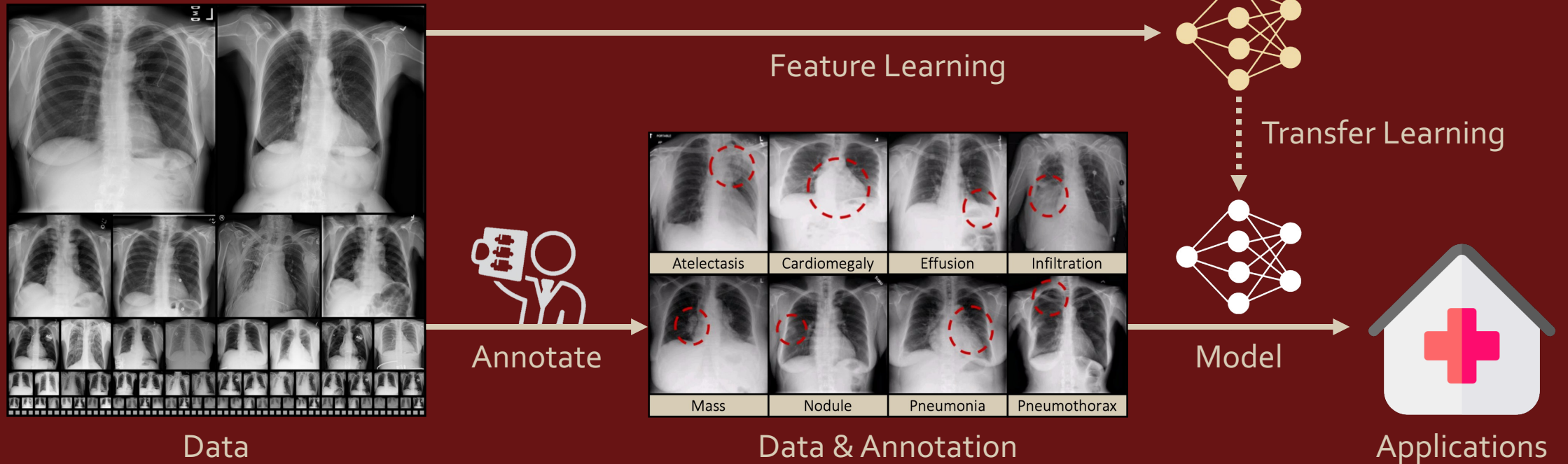
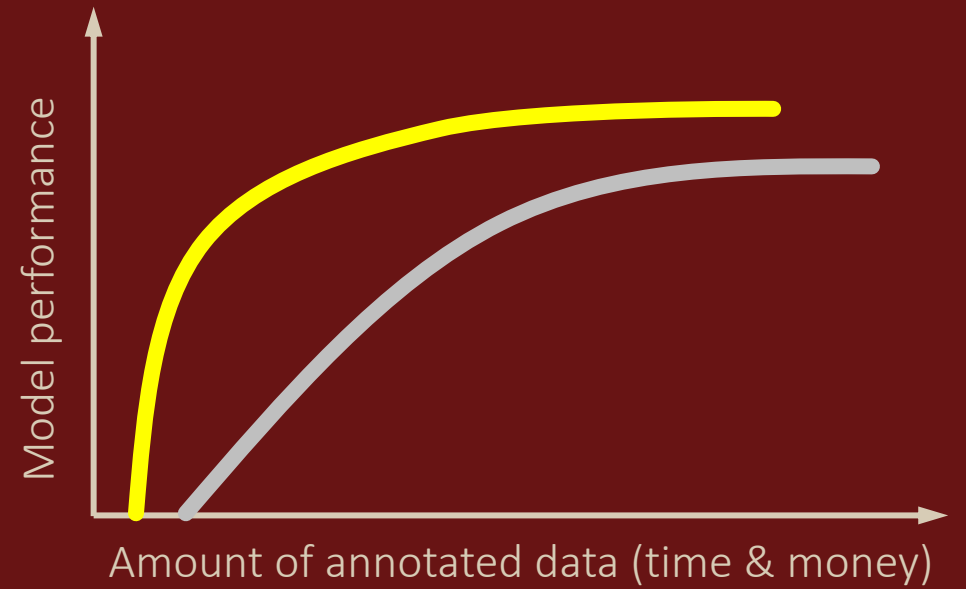


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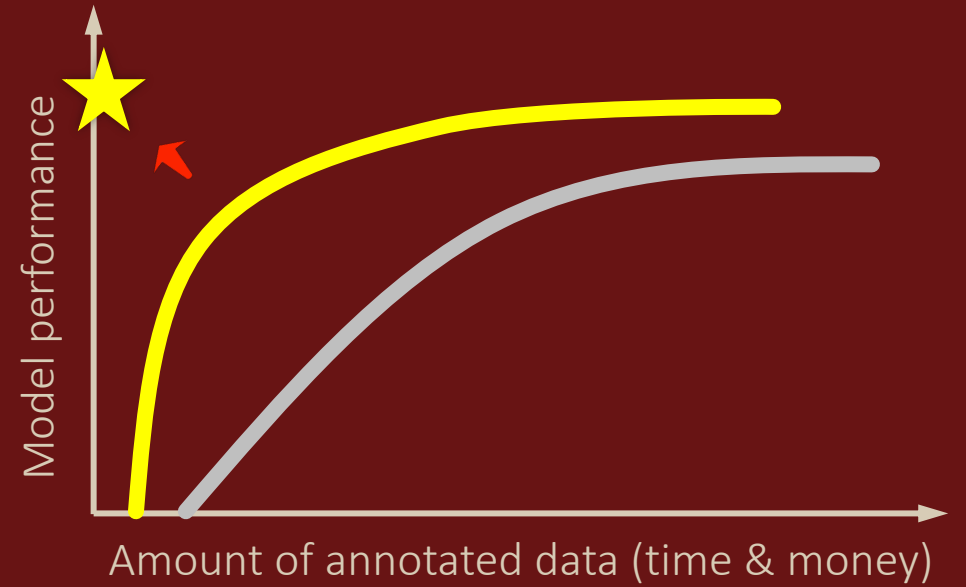
— PhD dissertation (*AMIA Doctoral Dissertation Award*) —

- Annotation-intensive deep learning
- Annotation-efficient deep learning

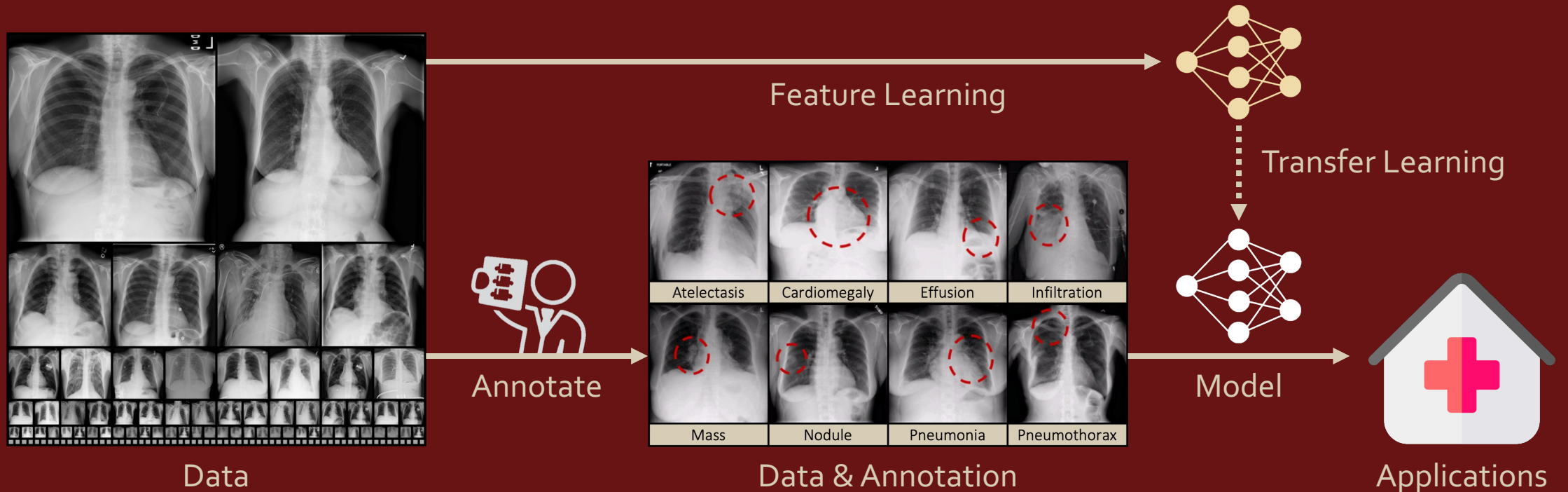


Goal: Minimize manual annotation efforts for rapid, precise computer-aided diagnosis systems.

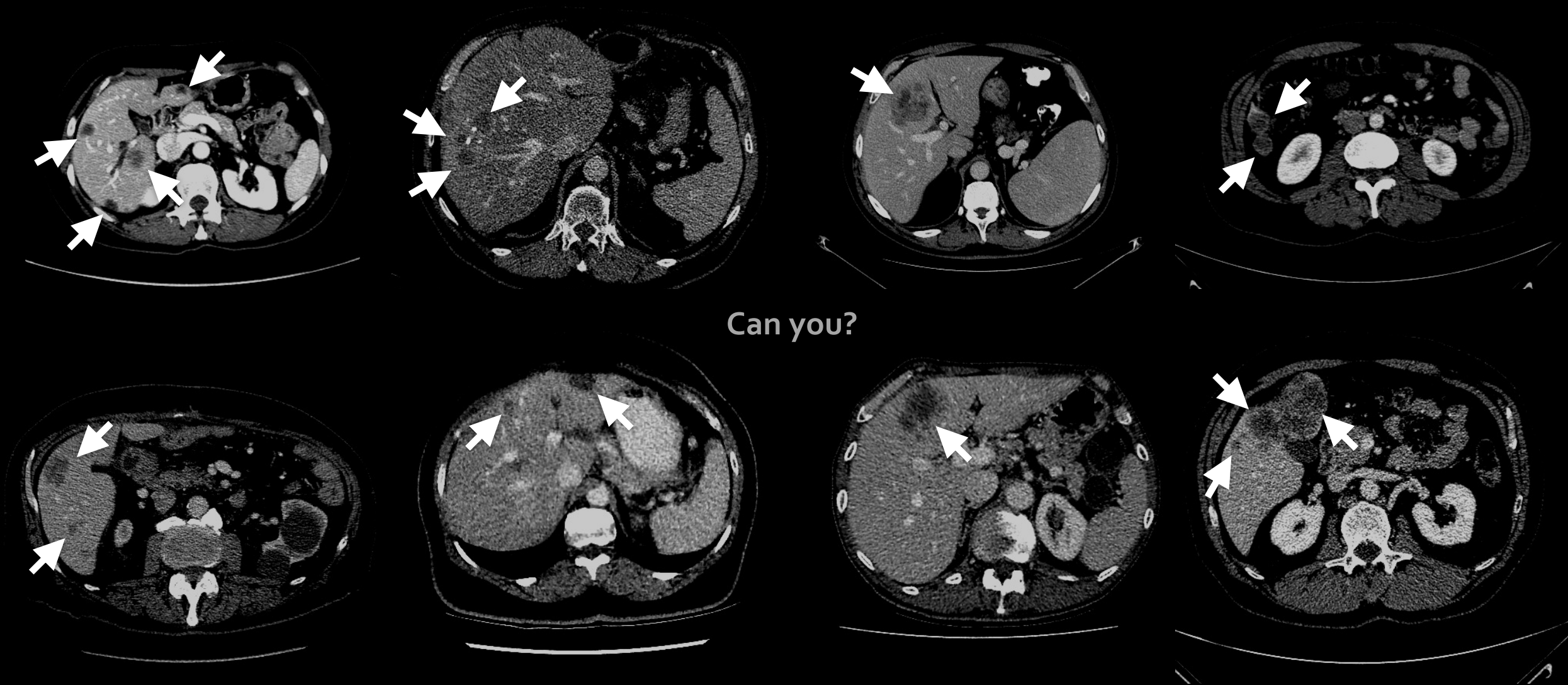
1. Acquiring necessary annotation efficiently from human experts.
2. Utilizing existing annotation effectively from advanced models.
3. Extracting generic knowledge directly from unannotated images.
—— PhD dissertation (*AMIA Doctoral Dissertation Award*) ——
4. Exploring ultra-weak annotation (radiology reports, synthetic data).



★ Annotation-free deep learning

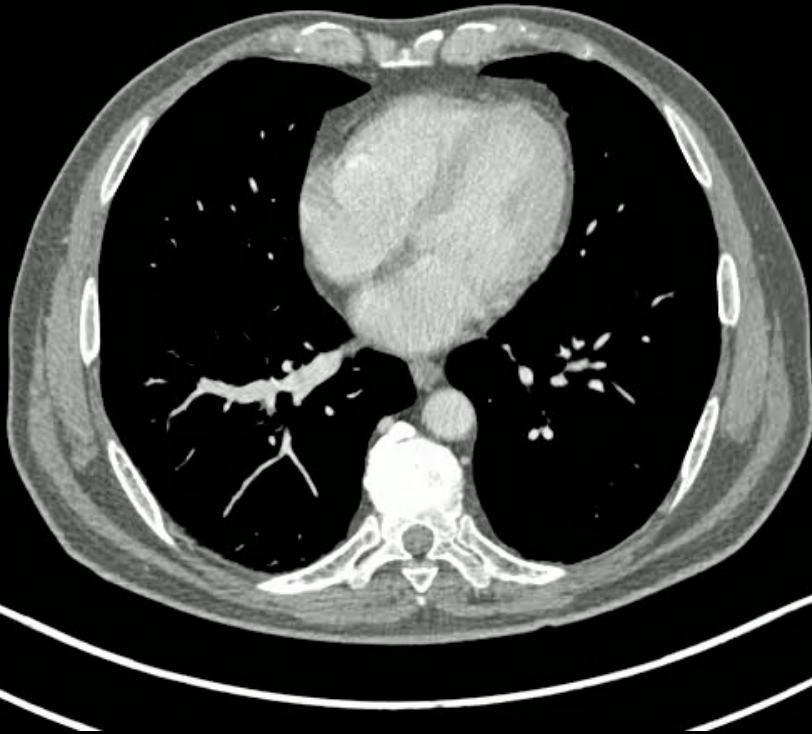


Medical professionals with over 6-year experience cannot tell which are real and which are synthetic tumors with an accuracy of 20% (lower than random guess)



Training AI on synthetic tumors performs almost as well as training it on real tumors.

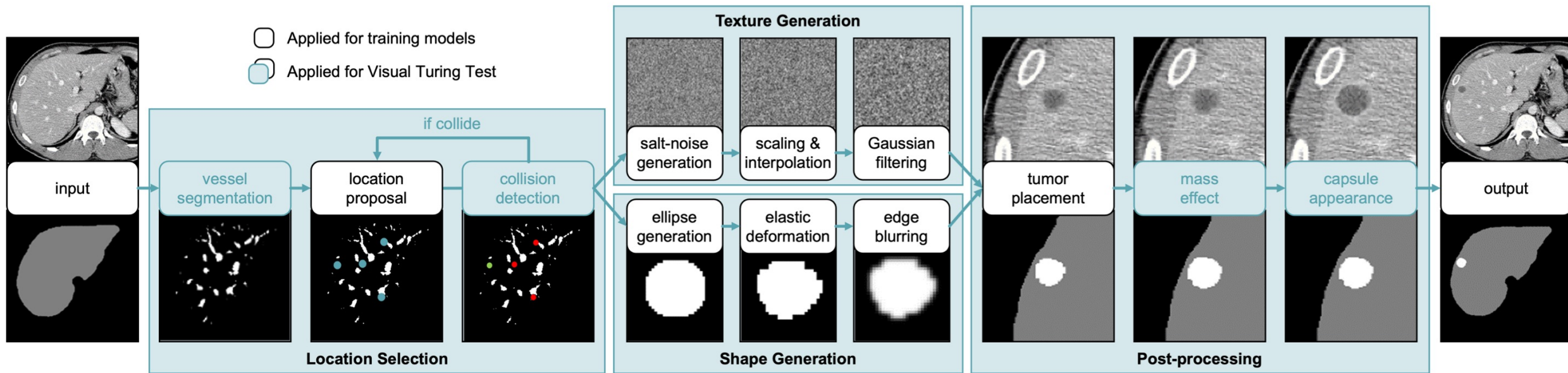
CT



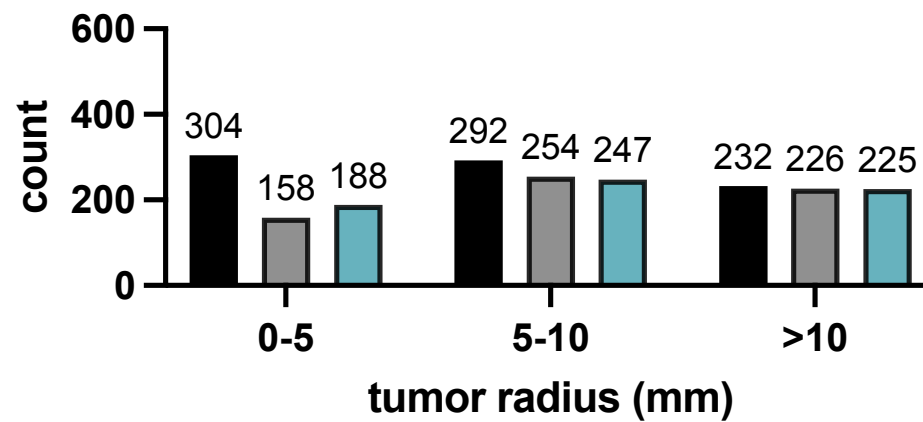
AI prediction
trained on real tumors
with per-voxel annotation
DSC = 58% [52% - 63%]

AI prediction
trained on synthetic tumors
with no annotation
DSC = 60% [55% - 65%]

- Liver
- Liver tumor



Code &
Awesome Tumor Synthesis



ground truth
 AI trained on real tumors
 AI trained on synt tumors

Significant

**Applications
Methodologies**

Impactful

**Annotation-intensive
deep learning**



**Annotation-efficient
deep learning**



**Annotation-free
deep learning**

**Earlier detection
of cancer**

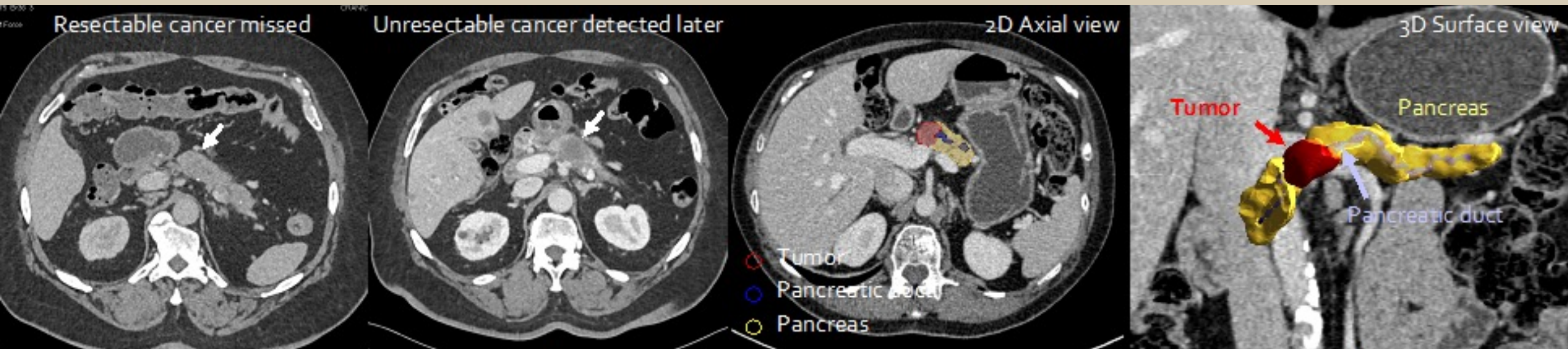
Significant

**Applications
Methodologies**

Impactful

Goal: Earlier detection of pancreatic neoplasms in CT scans using deep learning.

- 40,000,000 abdominal CT scans are performed each year in the United States.
- 1/3 of PDACs in these scans are missed by Radiologists. Early signs of PDAC can be subtle.
- Deep Learning can see things in images that most humans miss.
 - Pancreas tumors: 5,038 annotated CT scans in Johns Hopkins \blacksquare Sensitivity=97%, Specificity=99%
 - *This dataset took 15 years to annotate for a human.*



Goal: Earlier detection of pancreatic neoplasms in CT scans using deep learning.

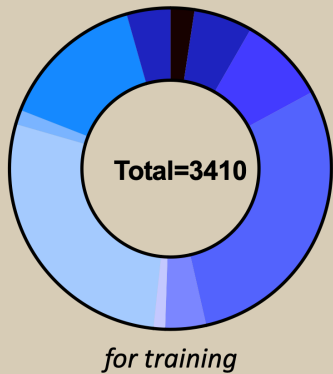
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Goal: Earlier detection of ~~pancreatic neoplasms~~ cancers in CT scans using deep learning.

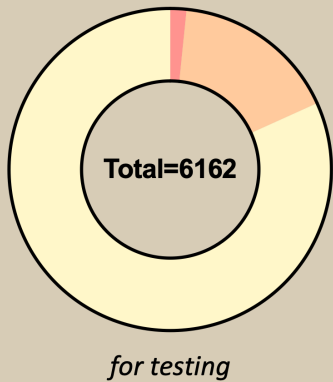
- Body Maps: This concept is similar to Google Maps, but it focuses on human anatomy rather than the Earth's geography. Body Maps offer several features:
 - (1) In-depth segmentation of anatomical structures.
 - (2) Disease screening across various structures.
 - (3) Language interaction between users and systems.

Two recently awarded grants: McGovern (\$400,000) and Lustgarten (\$1,922,421), where I served as Team Investigator.

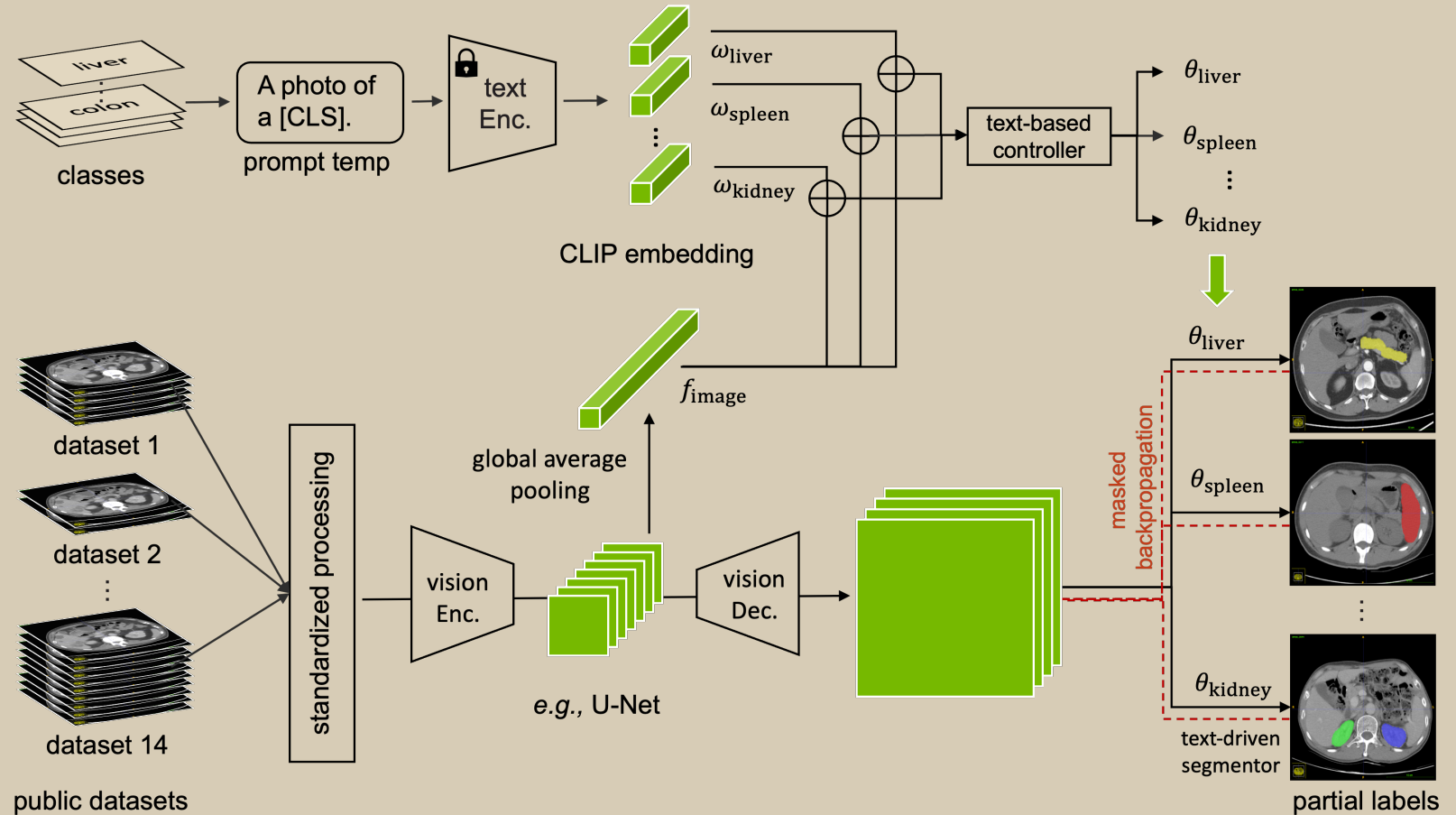
The first-place solution in Medical Segmentation Decathlon (MSD)



- 82 Pancreas-CT (1;0)
- 201 LiTS (1;1)
- 300 KiTS (1;1)
- 1000 AbdomenCT-1K (4;0)
- 140 CT-ORG (4;0)
- 40 CHAOS (4;0)
- 947 MSD (7;4)
- 50 BTCV (13;0)
- 500 AMOS (15;0)
- 150 WORD (16;0)



- 100 3D-IRCADb (13;0)
- 1024 TotalSegmentator (104;0)
- 5038 JHH (21;0)



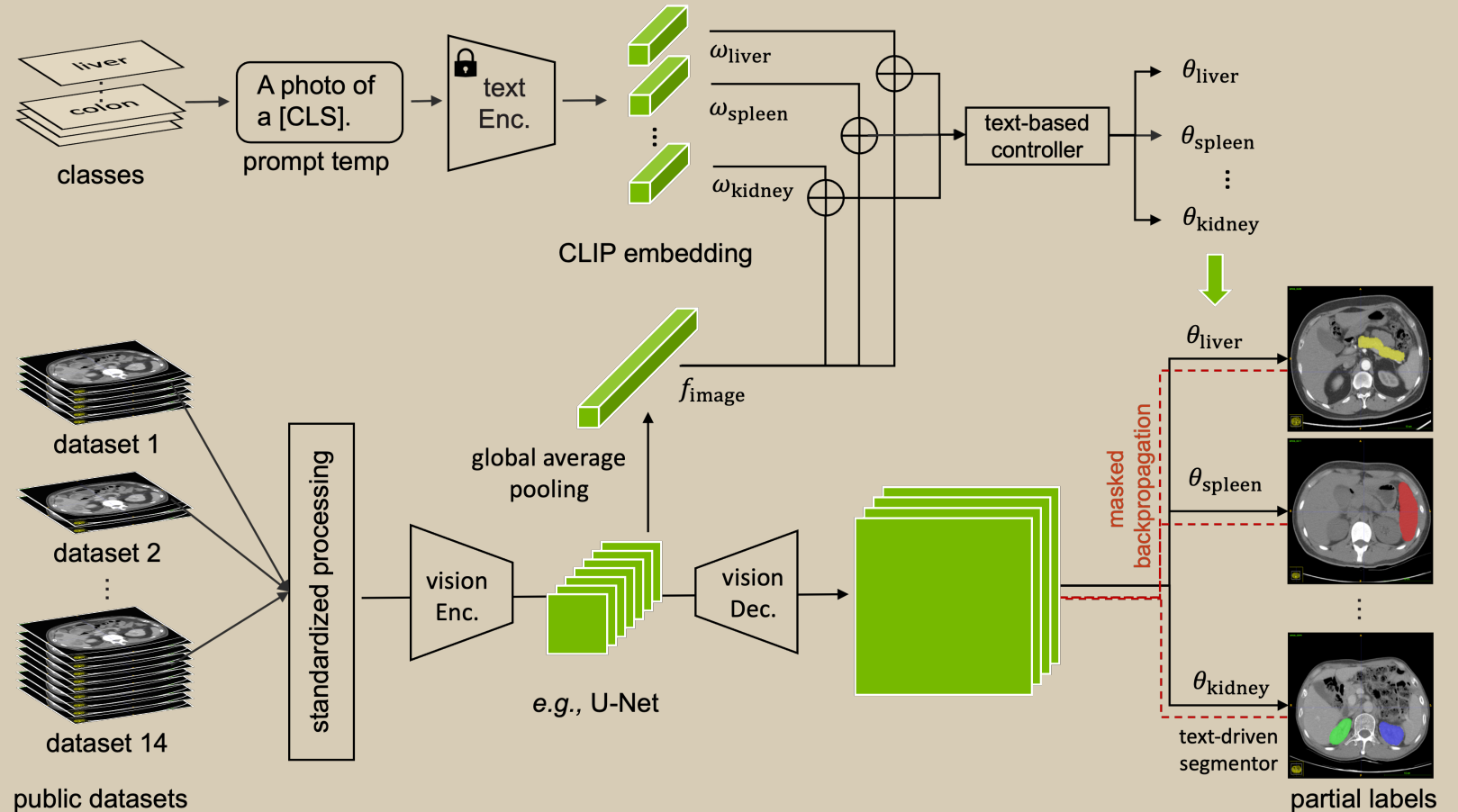
Liu, Jie, Yixiao Zhang, Jie-Neng Chen, Junfei Xiao, Yongyi Lu, Bennett A. Landman, Yixuan Yuan, Alan Yuille, Yucheng Tang, and Zongwei Zhou. "CLIP-Driven Universal Model for Organ Segmentation and Tumor Detection." arXiv preprint arXiv:2301.00785 (2023).

The first-place solution in Medical Segmentation Decathlon (MSD)



Code &
Awesome Medical Universal Models

25 Organs &
Six Types of Cancer
Kidney Tumor
Liver Tumor
Pancreas Tumor
Hepatic Vessel Tumor
Lung Tumor
Colon Tumor





Code &
Awesome Medical Universal Models

25 Organs &
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Info

Leaderboard

Statistics

Join

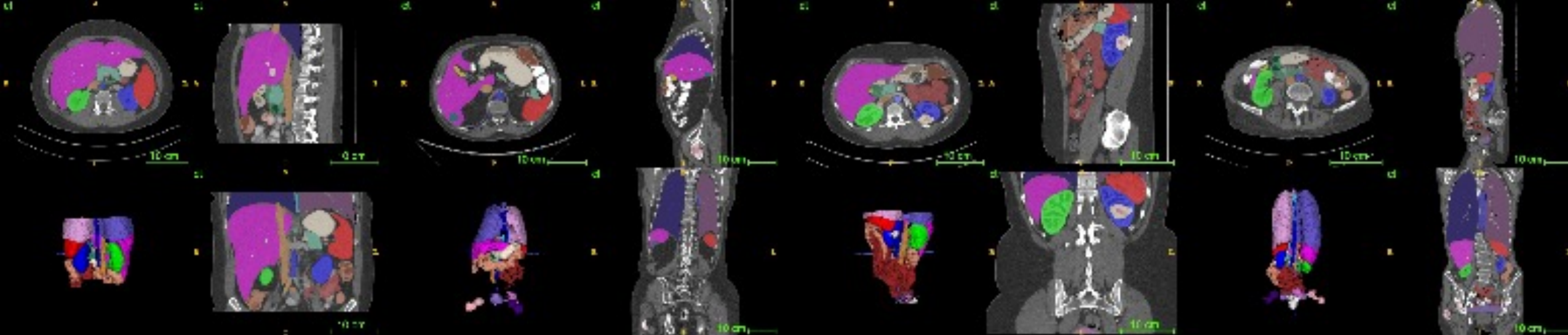
Challenge

Challenge Leaderboard

Search:

Additional metrics ▾ Show all metrics

#	↑ User (Team)	↕ Created	↕ Mean Position
1st	zongwei.zhou (universal_model)	13 Feb. 2023	5.6
2nd	Swin_UNETR Swin UNETR <i>CVPR-2022</i>	13 Nov. 2021	10.1
3rd	ahatamiz2	12 Nov. 2021	10.1
4th	lsensee nnU-Net <i>Nature Methods</i>	6 Dec. 2019	12.3
5th	AndyL	24 Nov. 2022	12.5
6th	heyufan1995 DiNTS <i>CVPR-2021</i>	30 Oct. 2020	12.8
7th	qsyung	5 Jan. 2023	12.9
8th	tangy5	11 Nov. 2021	13.5
8th	vishwesh.nath	11 Nov. 2021	13.5
10th	shrajanbhandary	7 March 2023	13.6

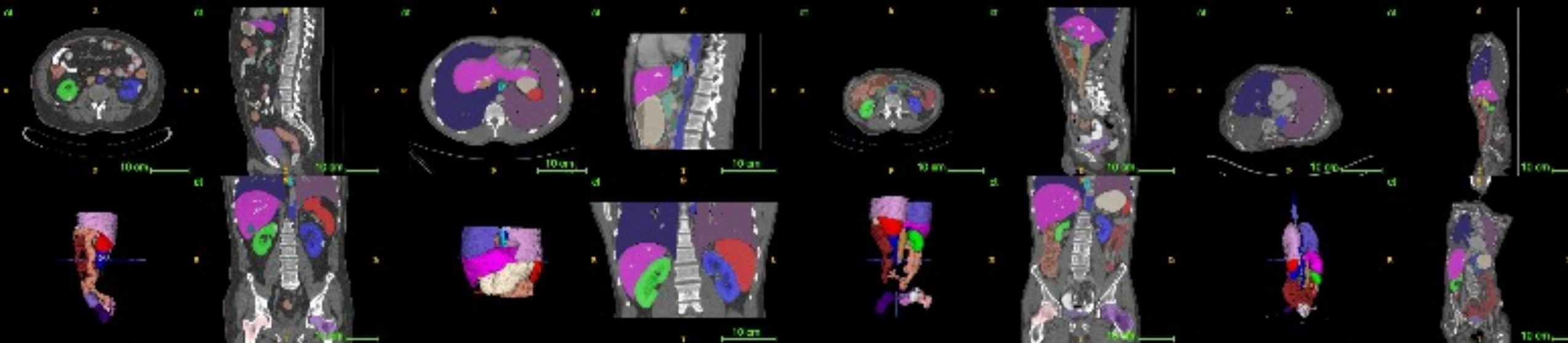


Pancreas-CT

LITS

KITS

AbdomenCT-1K

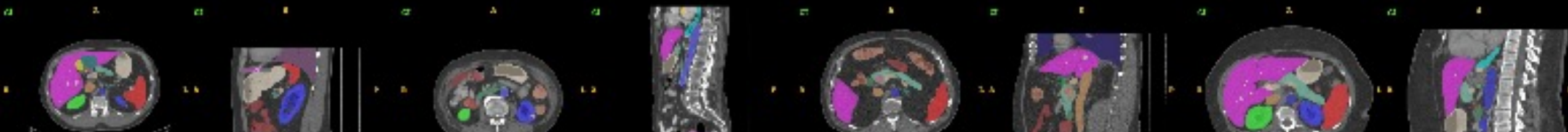


CT-ORG

CHAOS

MSD-Liver

MSD-Lung



Earlier detection
of cancers

Earlier detection
of cancer



Needful
Applications
Methodologies
Innovative



Annotation-intensive
deep learning



Annotation-efficient
deep learning



Annotation-free
deep learning

- 2021** **BMI 505: Foundations of BMI Methods II** Position: Teaching Assistant, Instructor: Dr. Sen Peng
- 2020** **BMI 598: NLP Methods for Biomedical Text Mining** Position: Teaching Assistant, Instructor: Dr. Murthy Devarakonda
- 2020** **BMI 598: Imaging in Diagnostics** Position: Teaching Assistant, Instructor: Dr. Jianming Liang
- 2019** **BMI 507: Intro Digital Image Processing** Position: Teaching Assistant, Instructor: Dr. Jianming Liang
- Possible topics: Medical Imaging Signals and Systems, Introduction to Bioengineering, etc.

Teaching Experience

Professional Service

Guest Editor:

- Journal of Imaging, Special Issue on “Imaging Informatics: Computer-aided Diagnosis”
- Applied Sciences, Special Issue on “Artificial Intelligence in Biomedical Image Processing”
- Machine Intelligence Research, Special Issue on “Multi-Modal Representation Learning”
- Frontiers in Radiology, Special Issue on “AI Applications for Cancer Diagnosis in Radiology”
- Sensors, Special Issue on “Advances of Deep Learning in Medical Image Interpretation”

Journal Reviewer: IEEE Transactions on Pattern Analysis and Machine Intelligence, IEEE Transactions on Neural Networks and Learning Systems, International Journal of Computer Vision, Medical Image Analysis, Artificial Intelligence in Medicine, Information Fusion, IEEE Transactions on Medical Imaging, Pattern Recognition, Computer Methods and Programs in Biomedicine, IEEE Transactions on Biomedical Engineering, Journal of Biomedical and Health Informatics, IEEE Access, Journal of Biomedical Informatics

Workshop Co-Organizer: ICML’23 and ICML’22 Workshop on Interpretable Machine Learning in Healthcare (IMLH)

Conference Program Committee: CVPR’22-23, ICCV’21-23, MICCAI’20-23, AAAI’20-23

Thank You!

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