



# BUILDING IN-HOUSE RADIOLOGY AI: Initial Results from the Pulmonary Embolism AI Initiative

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# AGENDA

Background

Methodology

Current Results

Project Progress

Conclusion

Acknowledgement

# BACKGROUND



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# CLINICAL IMPACT OF PULMONARY EMBOLISM

- Important **cause of in-patient mortality and morbidity**
- **Rapid deterioration** if untreated
- Effective **treatment available** :
  - *Anticoagulants*
  - *Thrombolytics*
  - *Catheter-based thrombectomy*
  - *Surgical thrombectomy*
- Requires urgent diagnosis by **CT Pulmonary Angiogram (CTPA)** to initiate treatment for good patient outcome
- Increasing urgent CTPA and other imaging requests poses a **clinical dilemma** for both **Radiologists** and **Clinicians**

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# CLINICAL IMPACT OF PULMONARY EMBOLISM

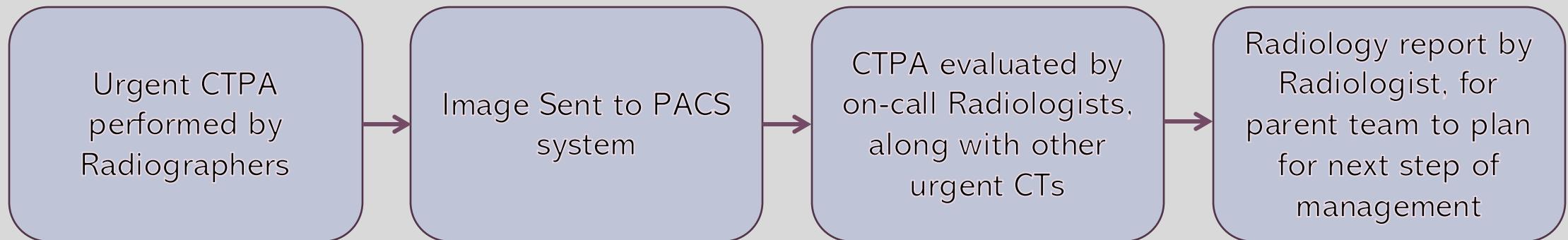
## The Pulmonary Embolism AI Initiative (Aims) :

1. To create an in-house AI pipeline for interpreting CT images to improve patient care.
2. The final AI solution shall screen CTPAs with a high negative predictive value (NPV) and detect right heart strain, flagging positive cases for immediate review and treatment.
3. Additionally, it shall provide explainable predictions for manual verification by Radiologists.

# CLINICAL IMPLEMENTATION OF AI

## - TO TACKLE THE CLINICAL DILEMMA

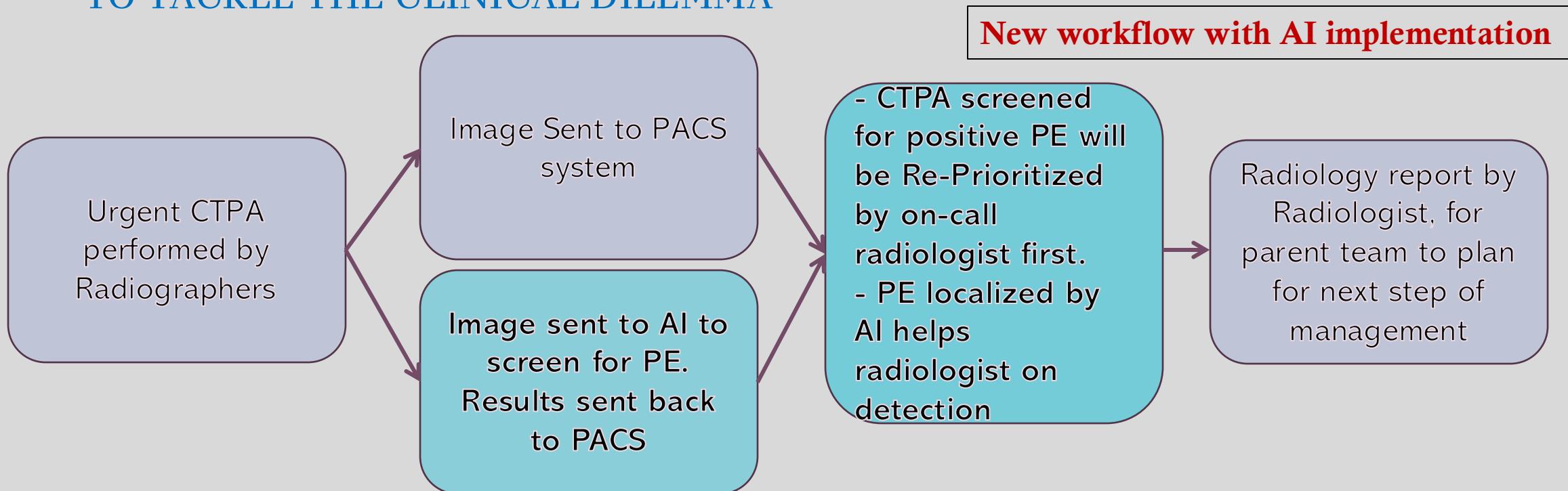
### Original Workflow (Without AI)



- Immense on-call workload
- Positive CTPA may be read late after studying other urgent CTs and may delay treatments
- Long report turn-around-time : Radiologists not only has to detect PE from CTPA, but also go through the entire CT thorax and upper abdomen for incidental findings

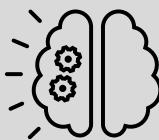
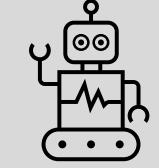
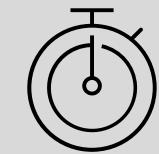
# CLINICAL IMPLEMENTATION OF AI

## - TO TACKLE THE CLINICAL DILEMMA



- Improve turn-around-time for positive PE → early treatment
- Improve accuracy (improved detection by AI support)

# VALUES AND PURPOSES



## Early Detection

- Automatic screening of positive CTPA during busy on-call hours
- Re-prioritize higher-risk cases
- Reduce reporting turn-around time for positive cases
- Enable early treatment**

## Diagnostic Support

- Enhance accuracy with AI-assisted detection
- Increase reading speed and confidence with highlighted findings (AI explainability)

## Clinical Research Opportunity

- Review prognostic factors for PE and Wells' criteria in Chinese Population

## Data Mining

- Generate service insights for future AI development and workflow optimization.

## Capacity Building

- For HA AI section and Radiologists on advanced imaging AI development
- For Radiologists to understand logics and caveats of AI solutions

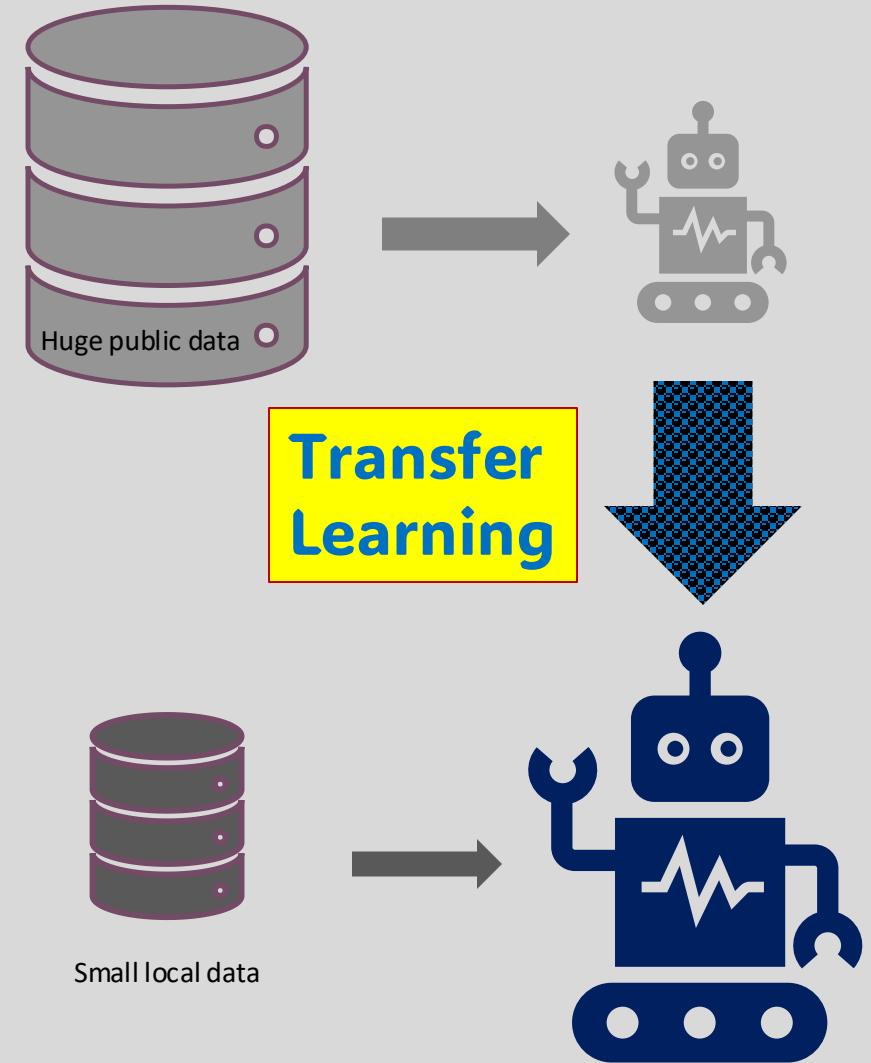
# METHODOLOGY



# DATA SETS

**Robust methodology by leveraging a huge public dataset and fine tuning using a local HA dataset**

- Public Dataset – labelled by expert Chest Radiologists
  - RSNA Classification Dataset (9,446 studies, 2.3 million images)
  - RSNA Localization Dataset (14,865 images with bounding boxes)
- Local Data
  - 573 CTPA cases from QMH
  - 173 positive for PE
  - Used for fine-tuning and validation



# BASE MODELS

## Classification Model

- Adopted and modified RSNA 2020 challenge winning model
- Transfer learning to adapt to local population

## Localization Model

- YOLOv11 architecture
- Provides bounding boxes for explainability

# RESULTS

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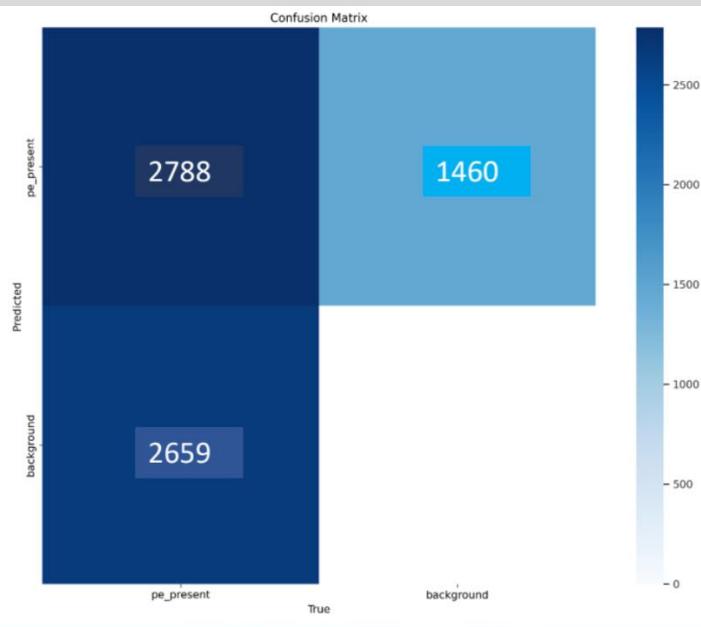
# TRAINING AND VALIDATION RESULTS

## Performance of Classification Model (RSNA Datasets)

- AUC of 0.952 on RSNA validation dataset
- AUC of 0.920 on representative validation set

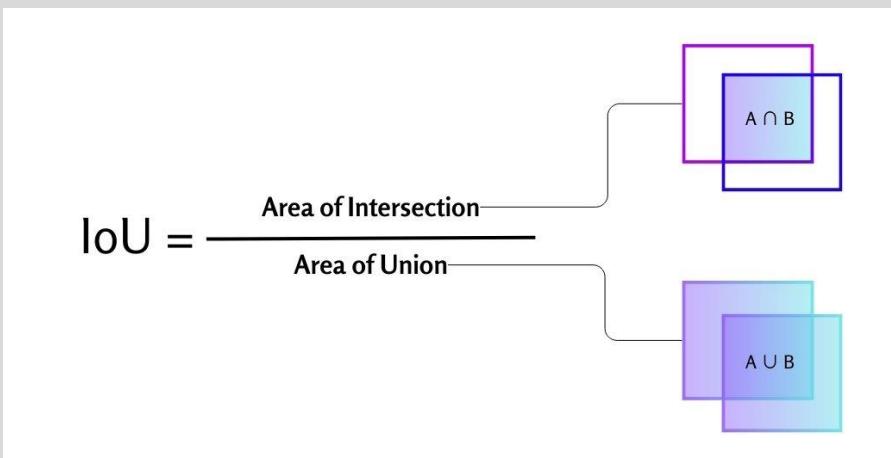
	Training	Ratio of pos/neg	Validation	Ratio of pos/neg	Val Loss	AUC
1	~1,431k	5:95	~359k	5:95	0.207	0.556
2	~154.4k	50:50	~38.6k	50:50	0.277	0.952
3	<b>~154.5k</b>	<b>50:50</b>	<b>~40.7k</b>	<b>5:95</b>	<b>0.394</b>	<b>0.92</b>

# TRAINING AND VALIDATION RESULTS



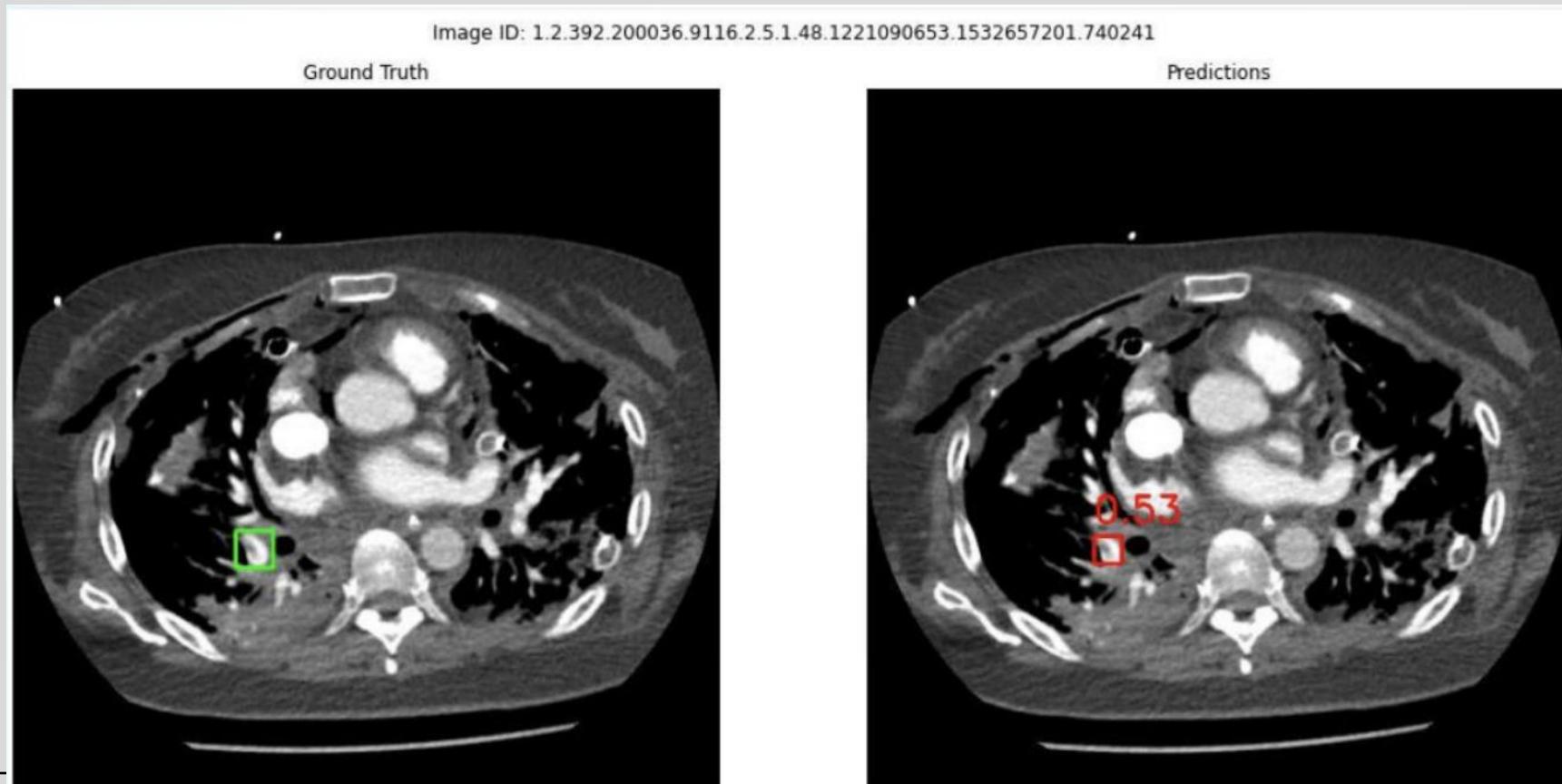
## Performance of Localization Model

- Precision: 0.52
- Recall: 0.47
- Accurately localized 69.2% of cases in local dataset testing
- Intersect over union (IOU) threshold : >0.5
- Successfully avoids mislabeling PE mimics



# PERFORMANCE ON LOCAL PE CASES (QMH)

## - True positives

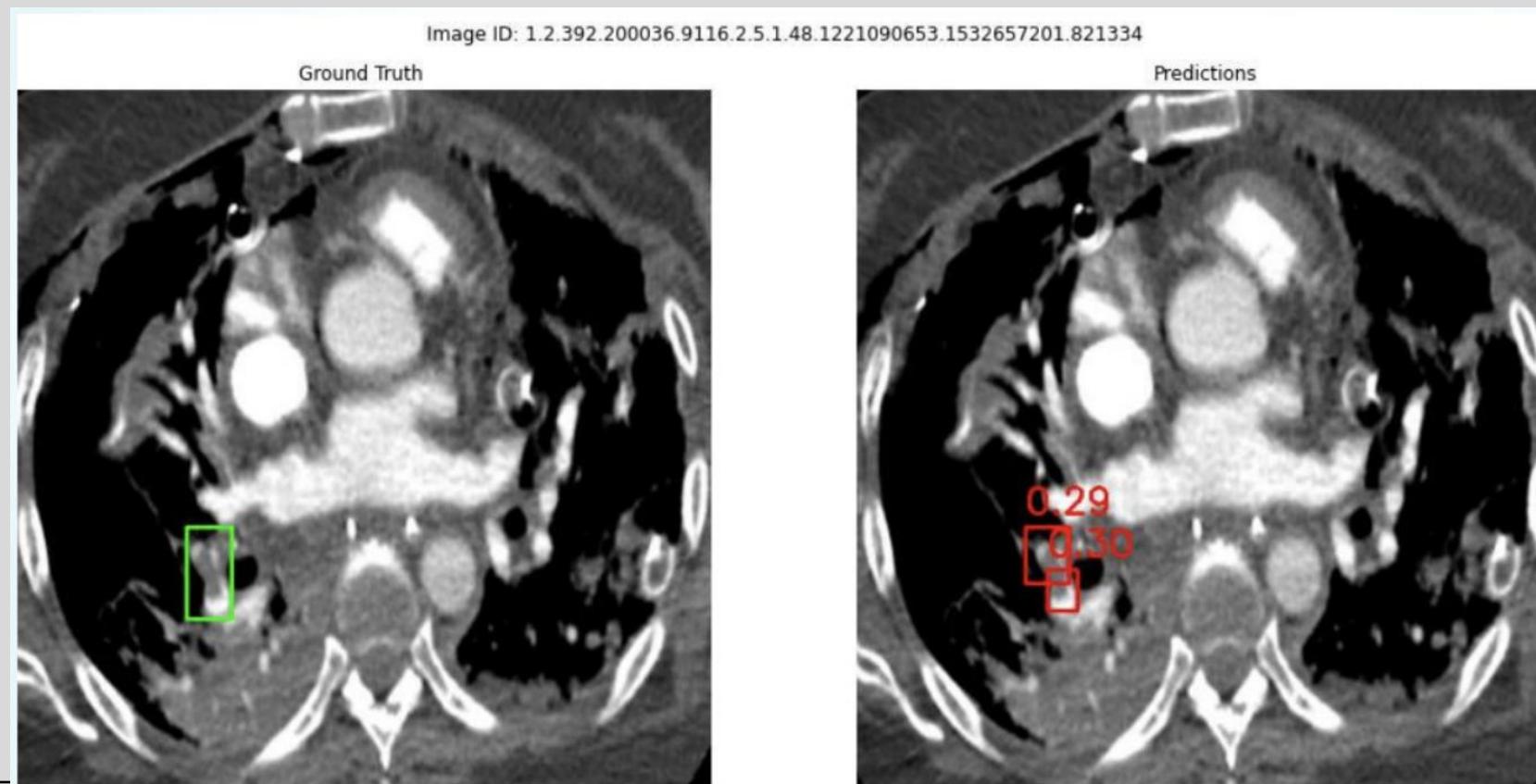


Doctor's Annotation

AI's Prediction

# PERFORMANCE ON LOCAL PE CASES (QMH)

## - True positives

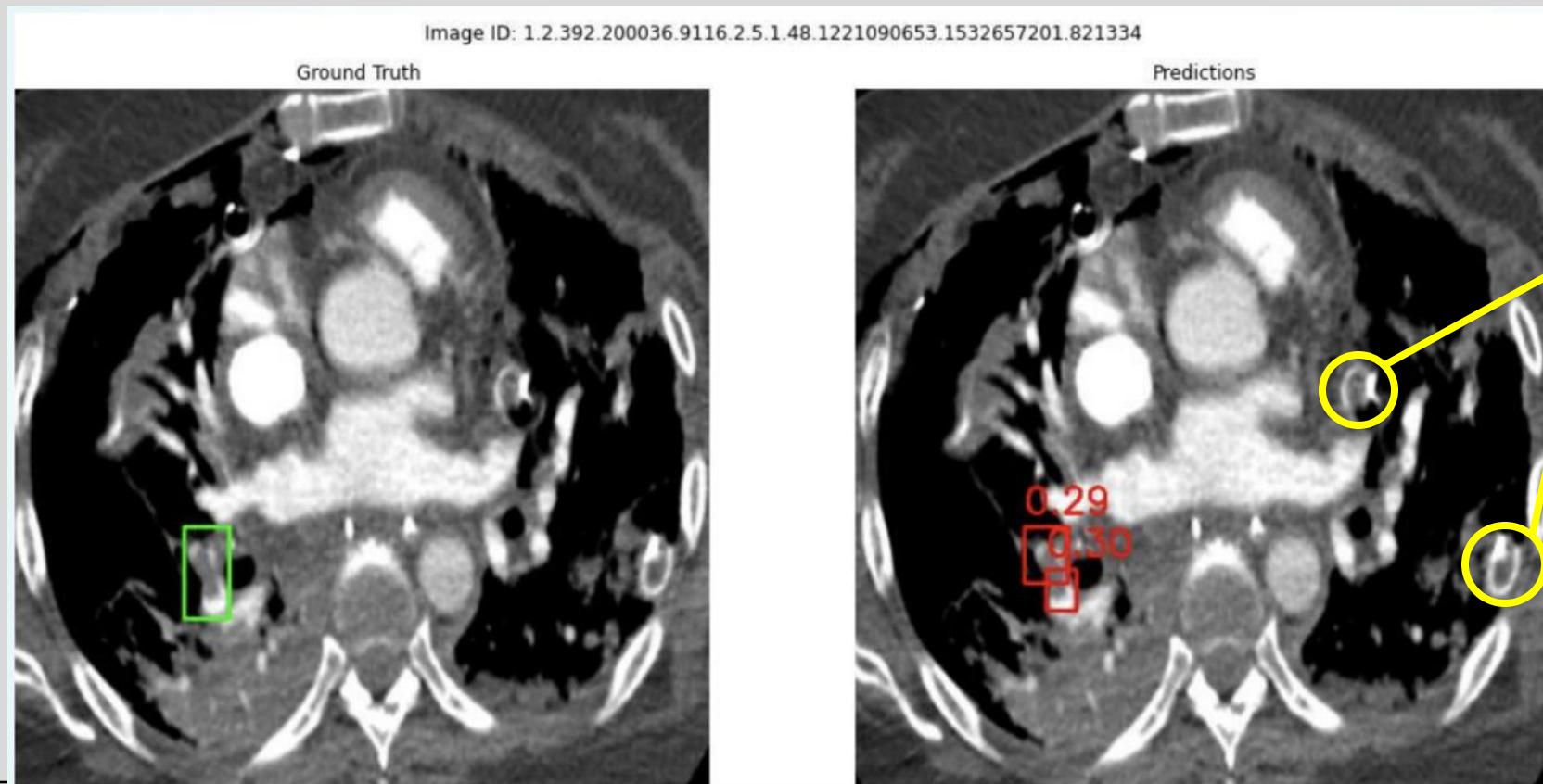


Doctor's Annotation

AI's Prediction

# PERFORMANCE ON LOCAL PE CASES (QMH)

## - True positives



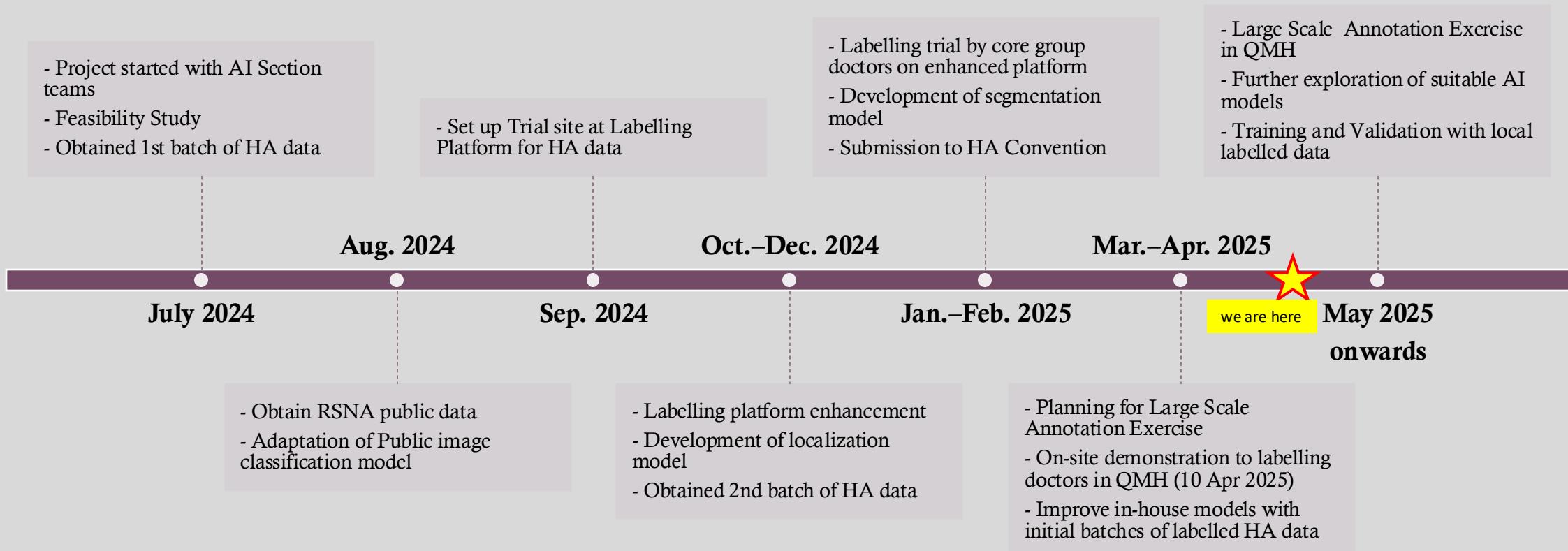
### Pericardial and Pleural drains

- Mimic filling defects of PE
- **AI avoided false positive prediction of PE mimics**

# PROJECT PROGRESS



# TIMELINE OVERVIEW



# NEXT STEPS

## Large Scale Annotation Exercise

- 573 QMH PE cases (173 positive) retrieved
- HA in-house labelling platform (AIAP) upgraded
- Large scale annotation exercise by QMH radiologists in May
- Including classification, localization and right heart strain annotation

## Model Development

- Local data will be used for training of current models through transfer learning
- Validation of base models on local dataset
- Development of segmentation model for RV/LV ratio

## Pipeline Development

- Continue development of a complete AI pipeline and final model
- Integrate models into clinical workflow
- Performance monitoring and feedback loop

## Scaling

- Pulmonary embolism is not an isolated problem in QMH
- Explore implementation in other Clusters
- Explore bigger scale, HA-wide annotation exercise for further training



# CONCLUSION

- **Feasibility of Local AI Development**

- First in-house CT AI solution at HA
- Demonstrates potential for clinical applications

- **Addressing Growing Demand**

- Utilizes advanced technology
- Collaborative effort between clinical and IT teams
- Support frontline doctors (Clinicians and Radiologists)

- **Improving Patient Outcomes**

- Early detection and treatment for better patient outcomes



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# ACKNOWLEDGEMENTS

## Project working group:

- Chan JCY(1), Li AMK(1), Yu SJA(1), Chan DLH(1), Lui KWC(2), Garifullin A(2), Lam YF(2), So BCH(2), Chung APM(2), Lee DPK(2), Chan JKY(2), Choi KH(1), Lam HY(1), Cheung CW(1)

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*Thank you very much!*