



Cloudphysician

VITAL EXTRACTION CHALLENGE

Inter IIT Tech Meet 11.0

PRESENTED BY
Team 55





Model Overview

Our proposed model has 3 main steps:

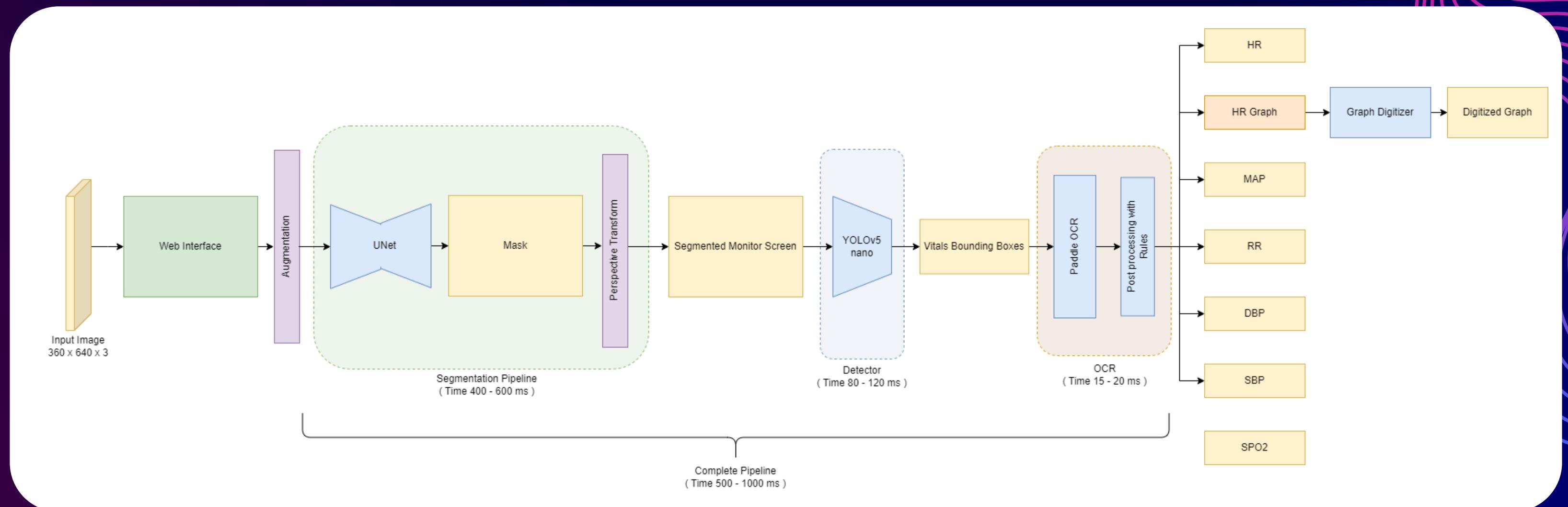
- **Step 1:** Segmentation
- **Step 2:** Detection
- **Step 3:** Character Recognition.

Inference time (on CPU): 0.6 - 1.0 s



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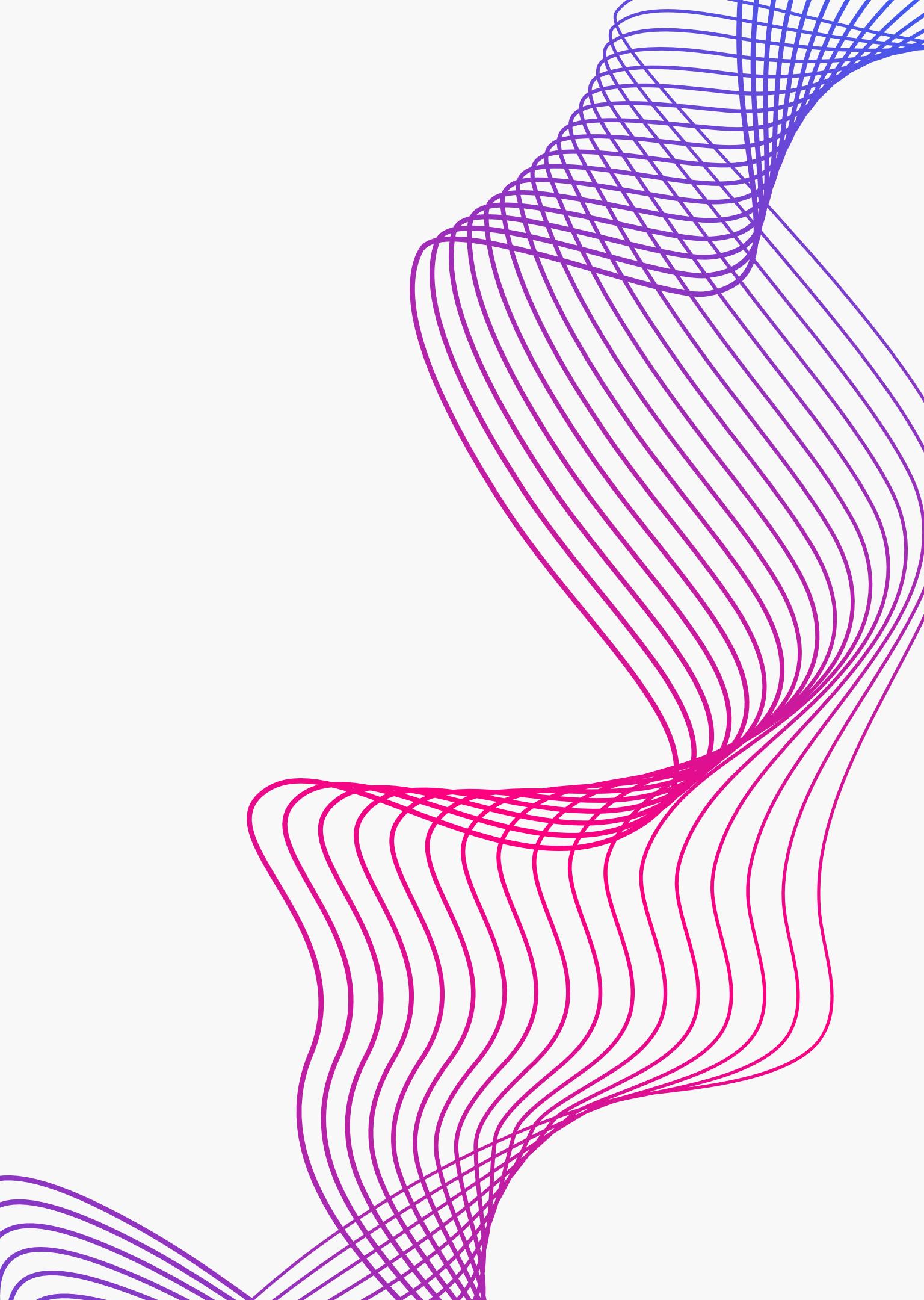
Proposed Model Pipeline



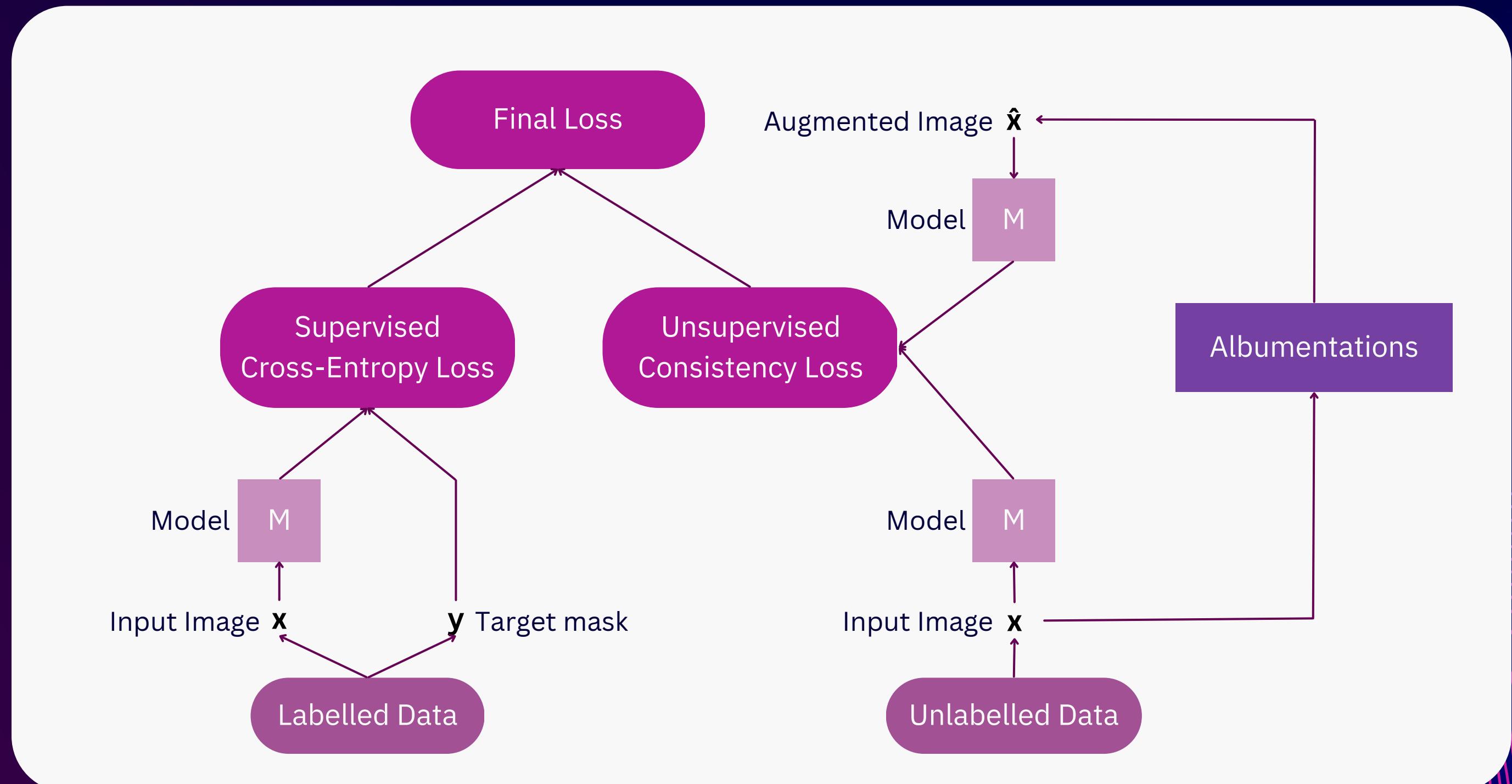


Step 1: Segmentation

- Segmentation based **U-Net architecture** (initialized with **IMAGENET weights**)
- **Semi-Supervised(SS) Semantic Segmentation Approach.**
- Novel additions: **LR scheduler, Augmentations variations**
(randAugment - a type of novel augmentation technique)



Semi-Supervised Segmentation

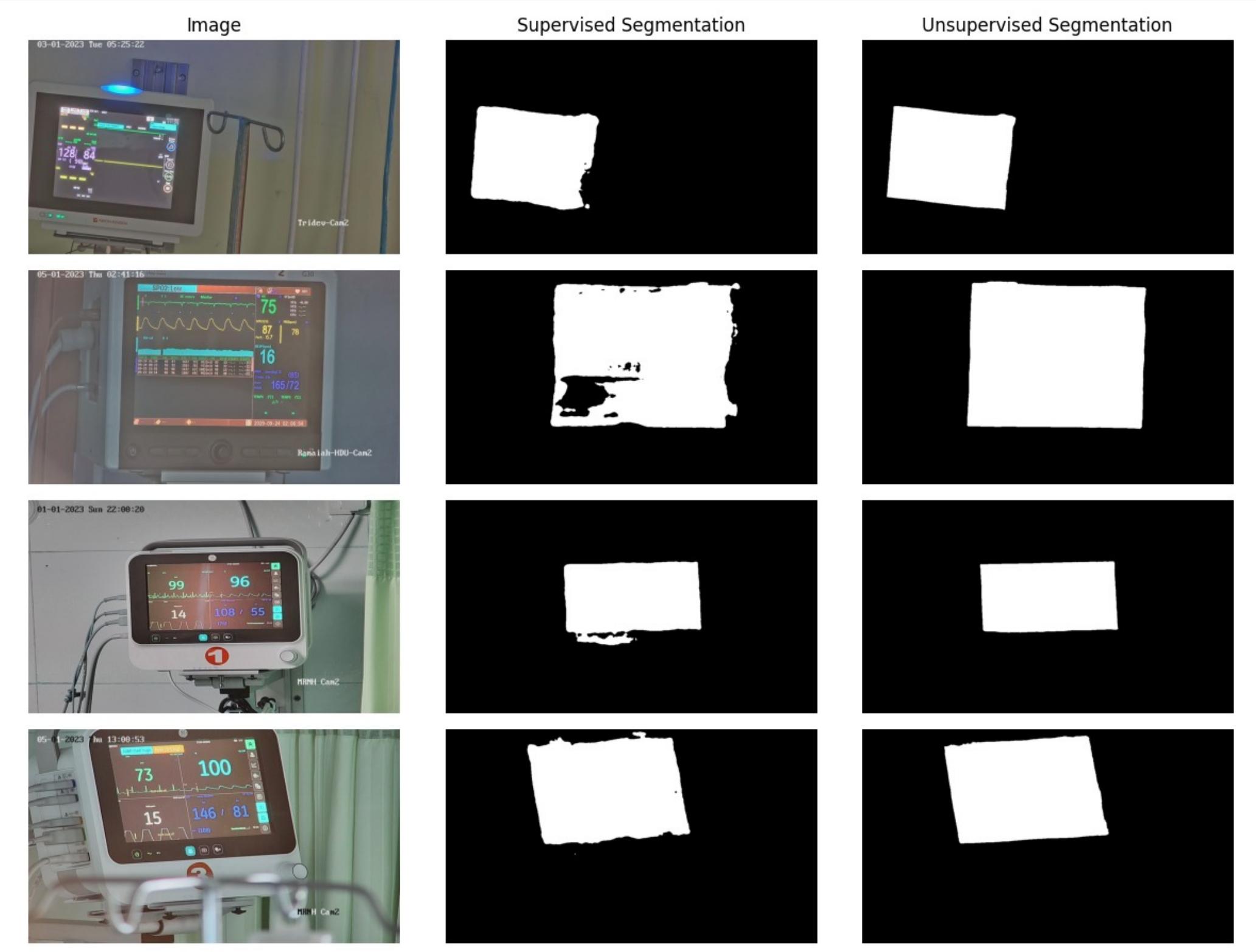




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Why choose Semi-Supervised Segmentation model?

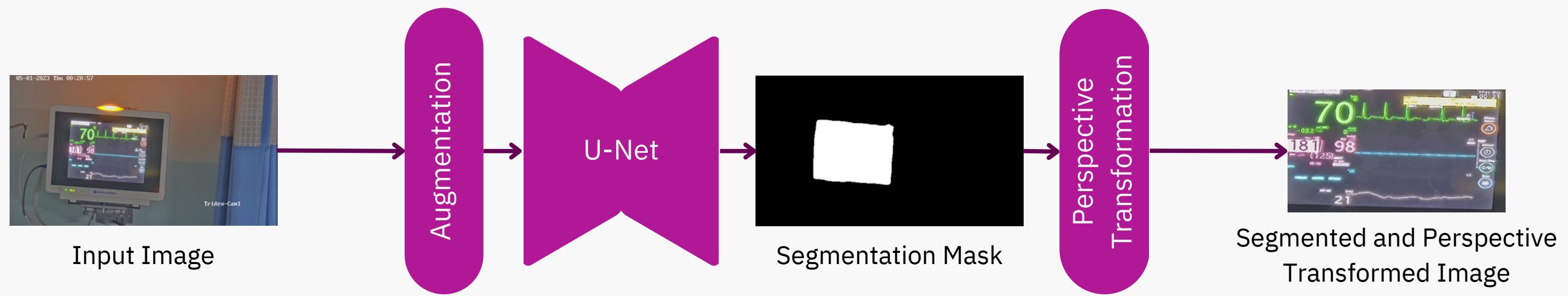
- monitor positions are not always rectangular
- detection model will not give good accuracy
- pixel level segmentation will perform better.





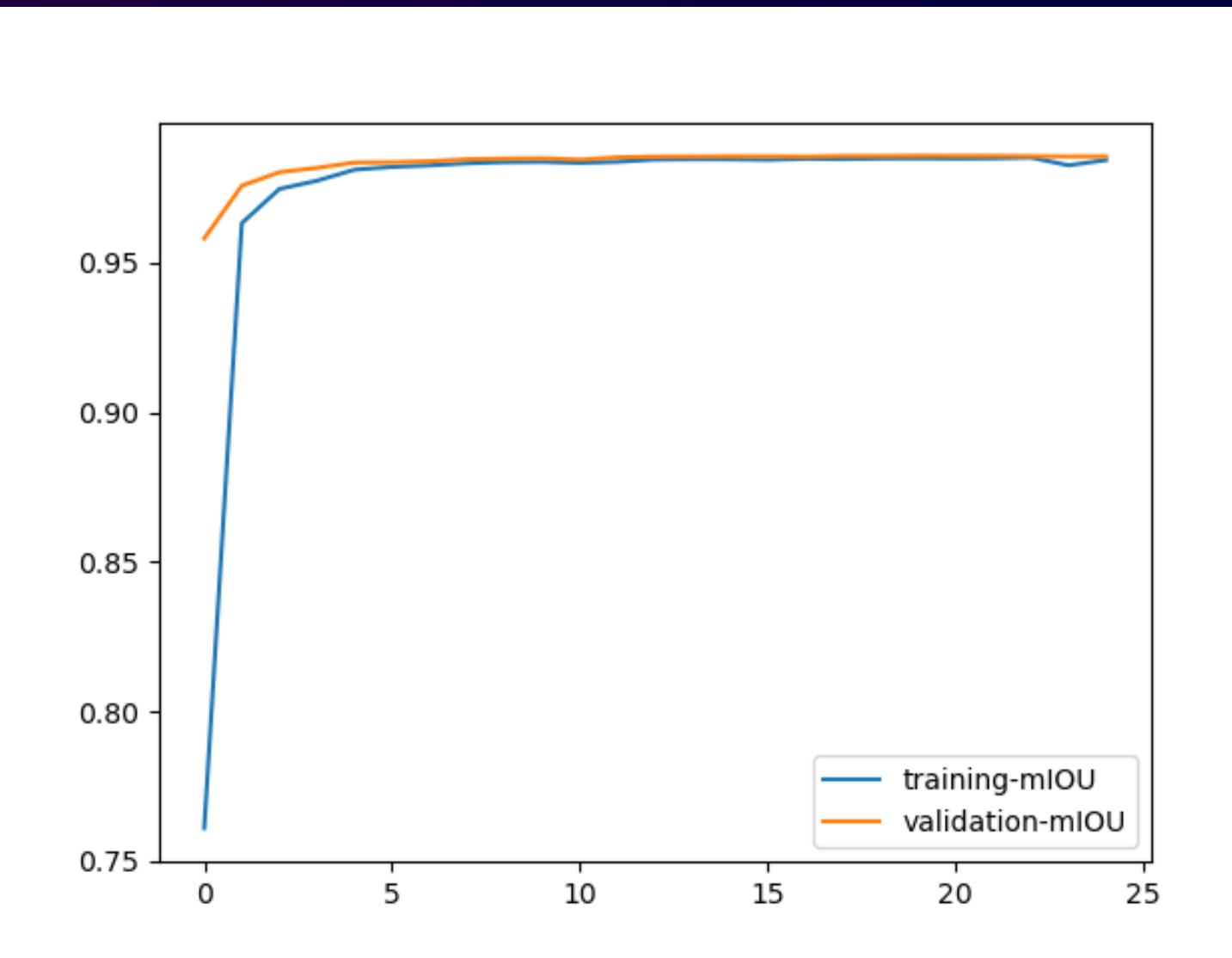
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Segmentation Pipeline

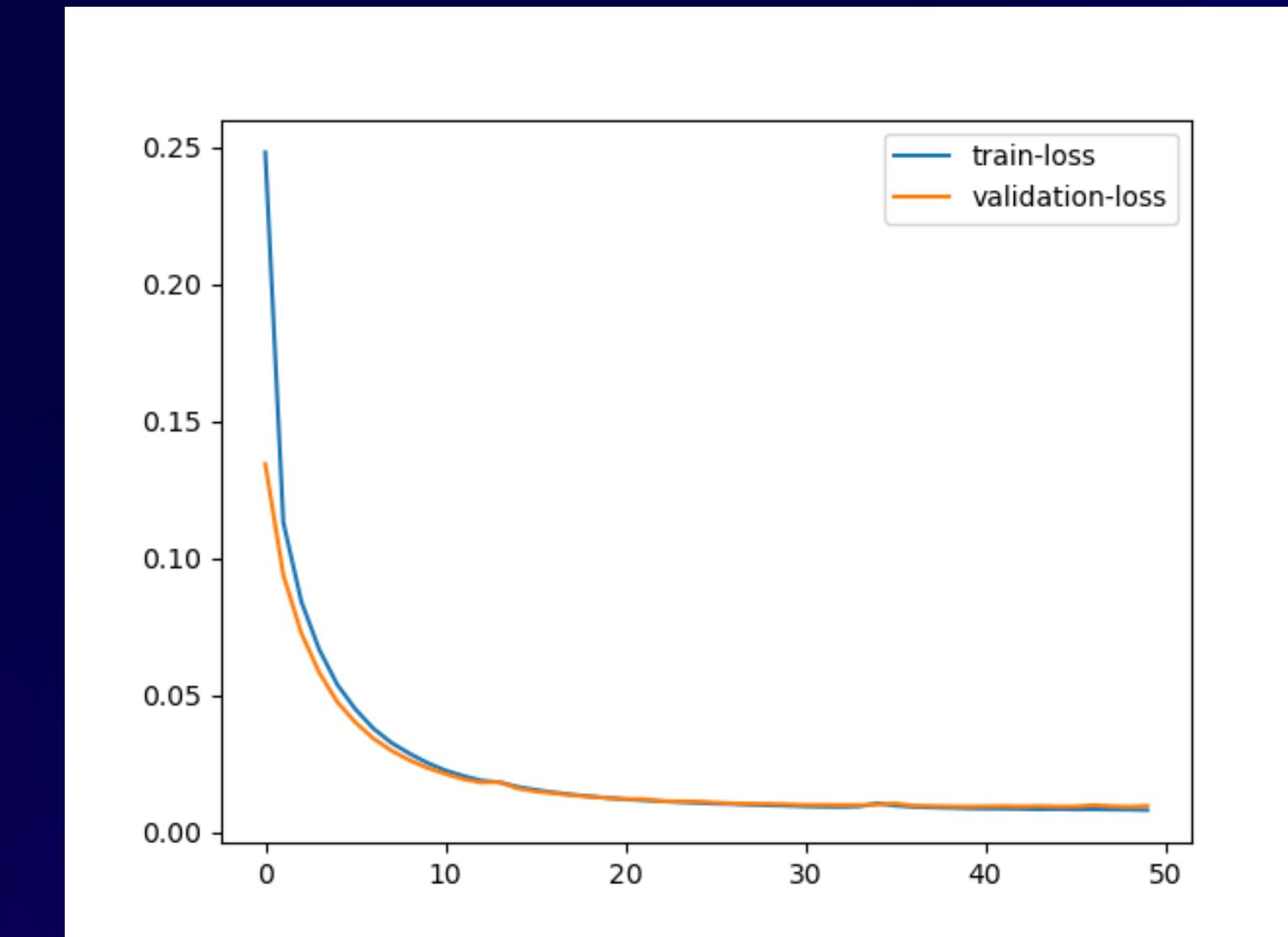


Inference time (on CPU): 0.4 - 0.6 s

Metrics



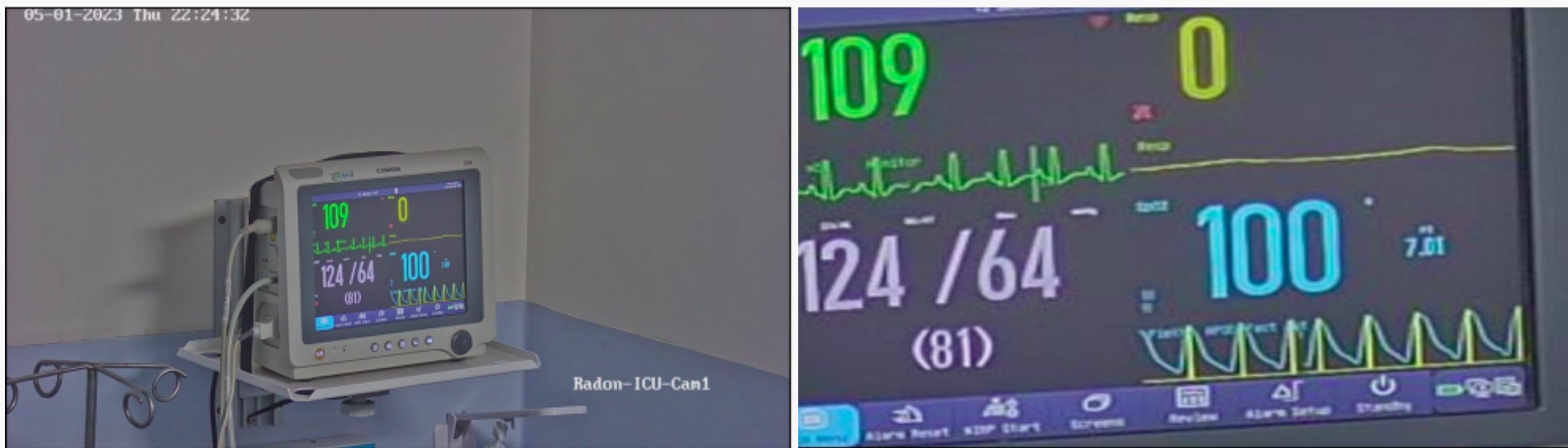
Plot for mIOU score



Plot for loss

Perspective Transformation

- **Convex hull** and **Contours methods** for finding mask corners.
- **OpenCV perspective transform** for wrapped perception.



Left - Original Image, **Right** - Segmented and Perspective transformed image



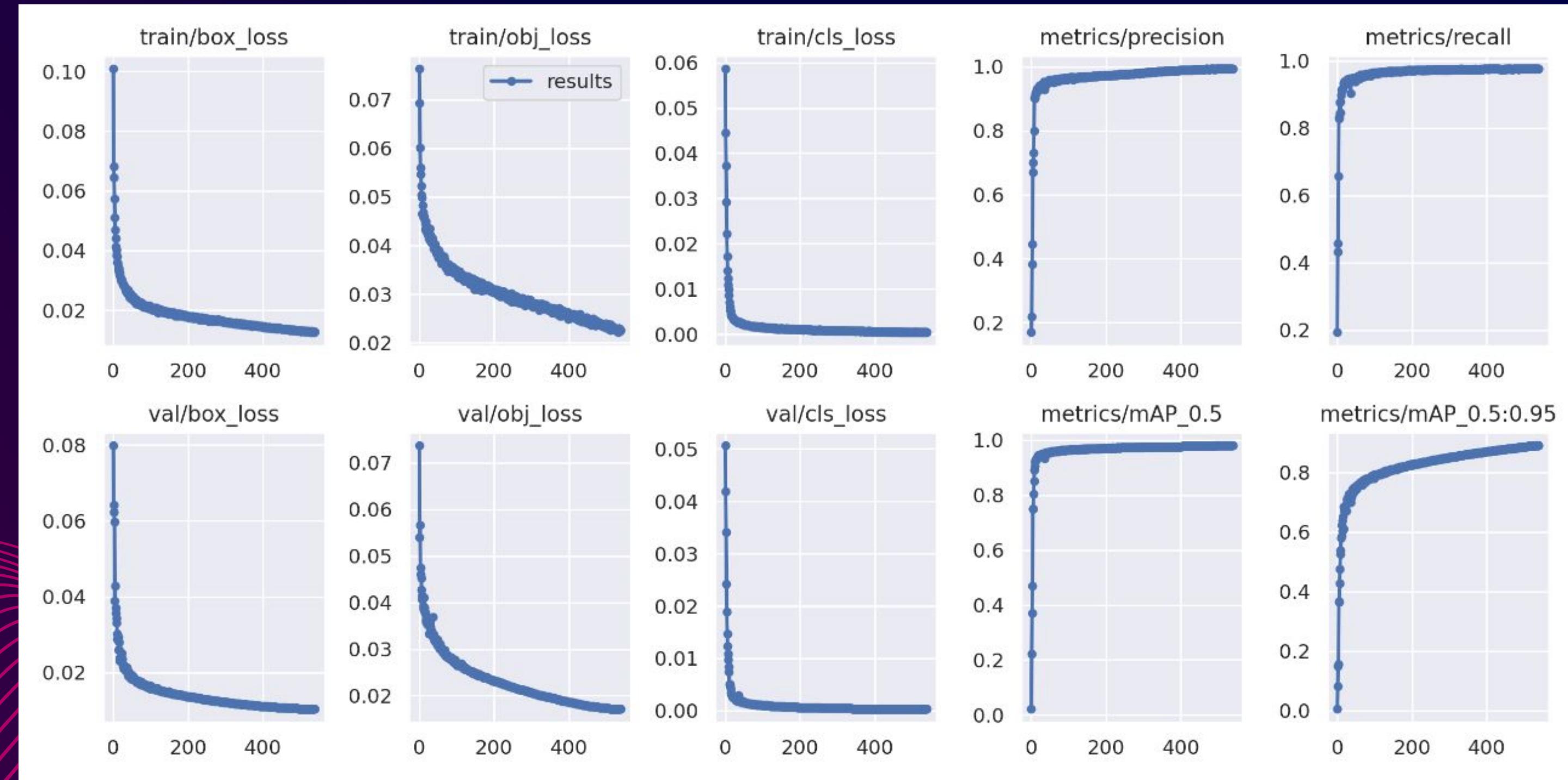
Step 2: Detection

- **YOLOv5nano** model - fast and lightweight
- Preprocessing and normalization done by the model itself
- **Inference time** (on CPU): 0.05 - 0.1 s



Top - Perspective Transformed Image
Bottom - Yolo Extracted Vitals and Graph

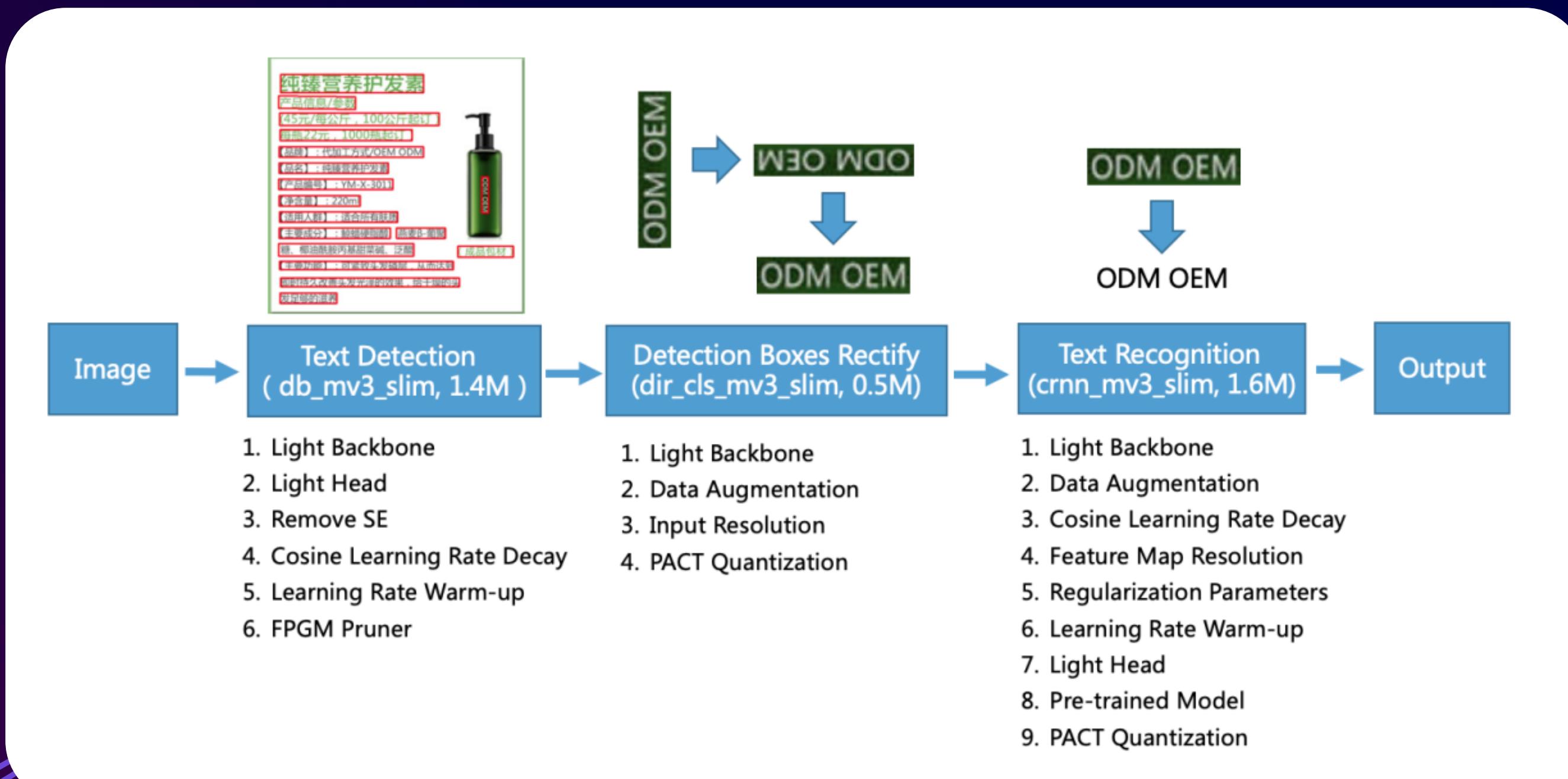
Metrics



Step 3: Character Recognition

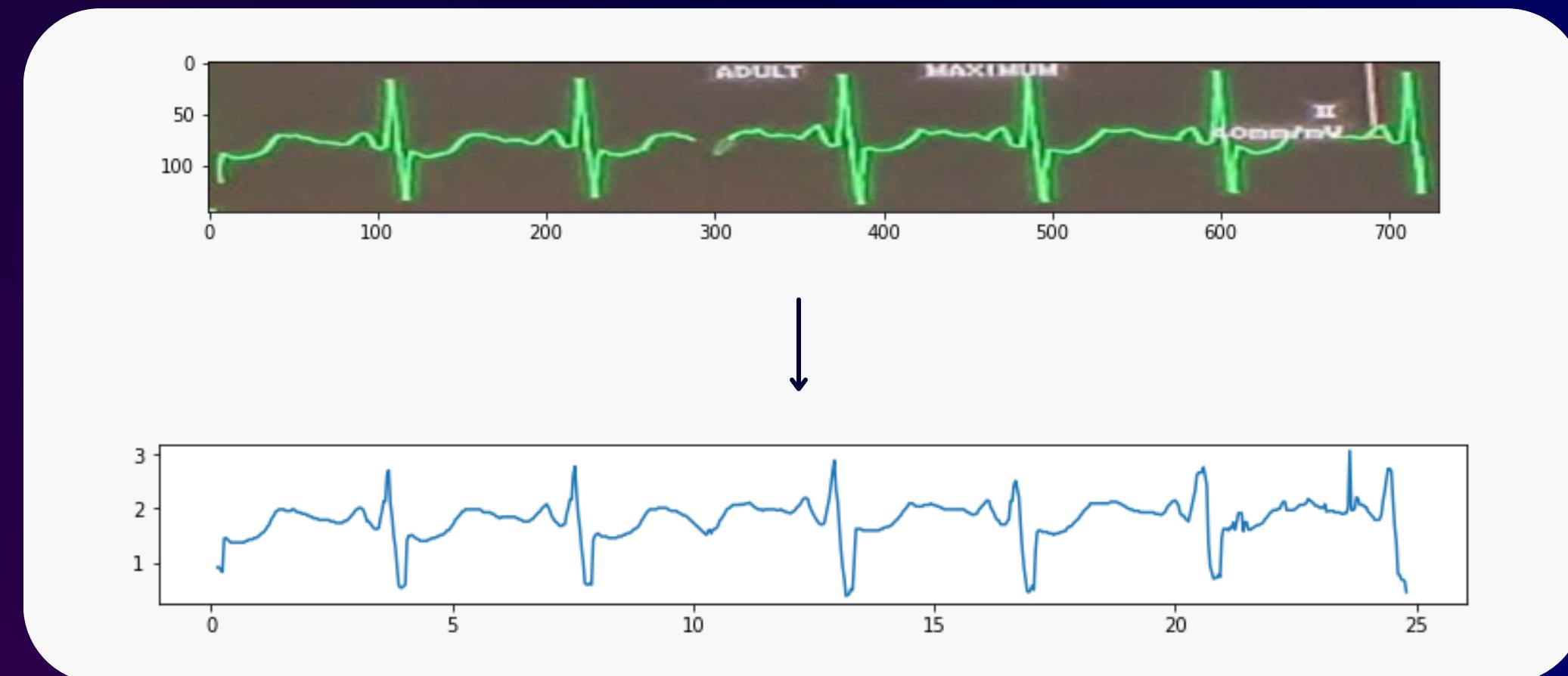
- **PaddleOCR**, an open source OCR model
- Performed **OCR recognition** for vitals from the bounding boxes extracted from YOLO.
- Done **post-processing** on the predicted results based on monitor layout info.
- **Inference time** (on CPU): 0.15 - 0.20 s

PaddleOCR Framework



Graph Digitizing

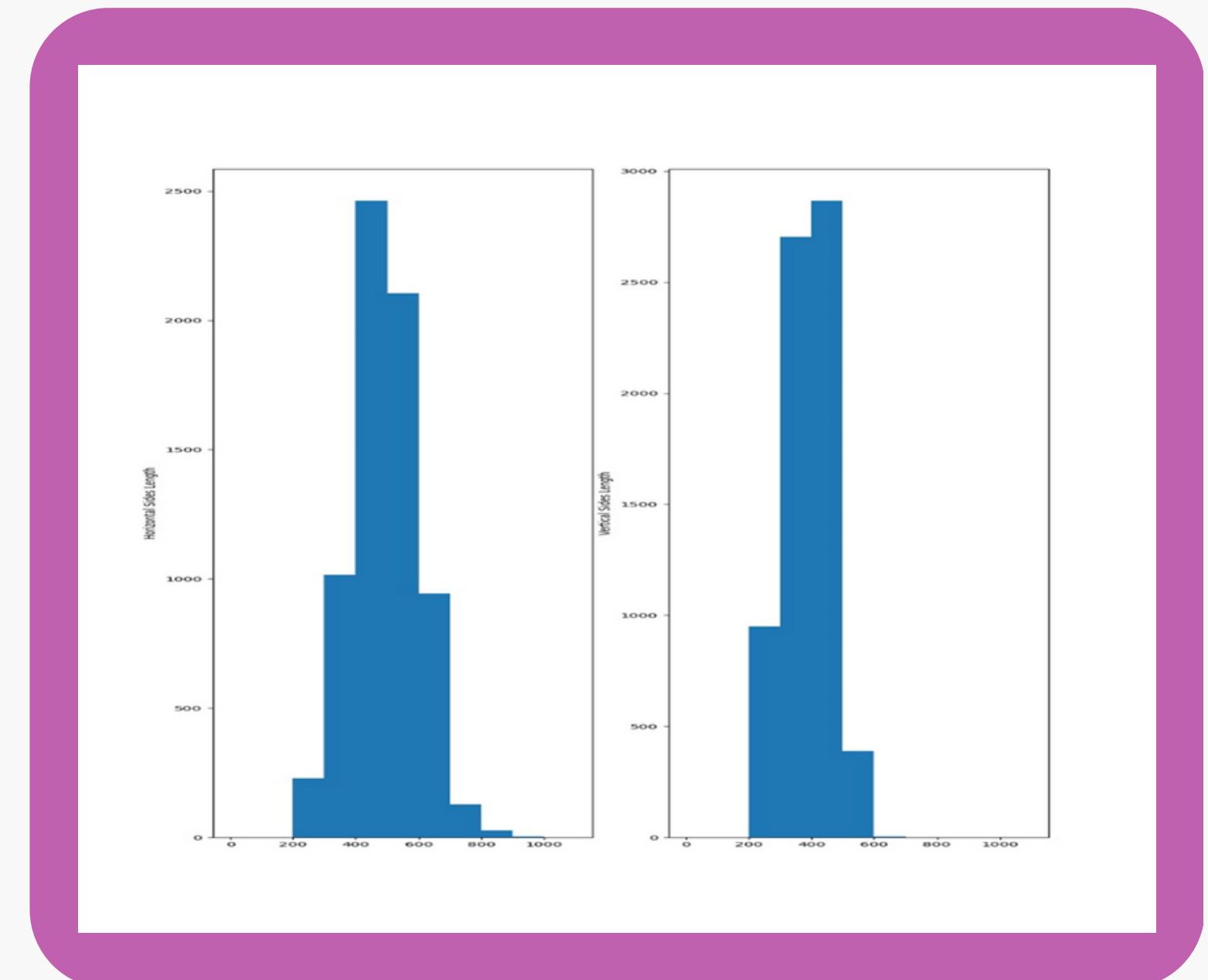
- Python implementation of **WebPlotDigitizer**.
- WebPlotDigitizer - web based tool to extract numerical data from plot images or maps.



Novelty

In terms of Speed

- **Multiprocessing** during OCR recognition boosted speed by **2.5X**.
- **YOLOv5nano** for direct extraction of vitals very fast on CPU
- **Carefully chosen input size** - plotted histogram of both dimensions of extracted masks and took the mean of histogram as the optimal dimension for YOLO.





Novelty

In terms of Accuracy

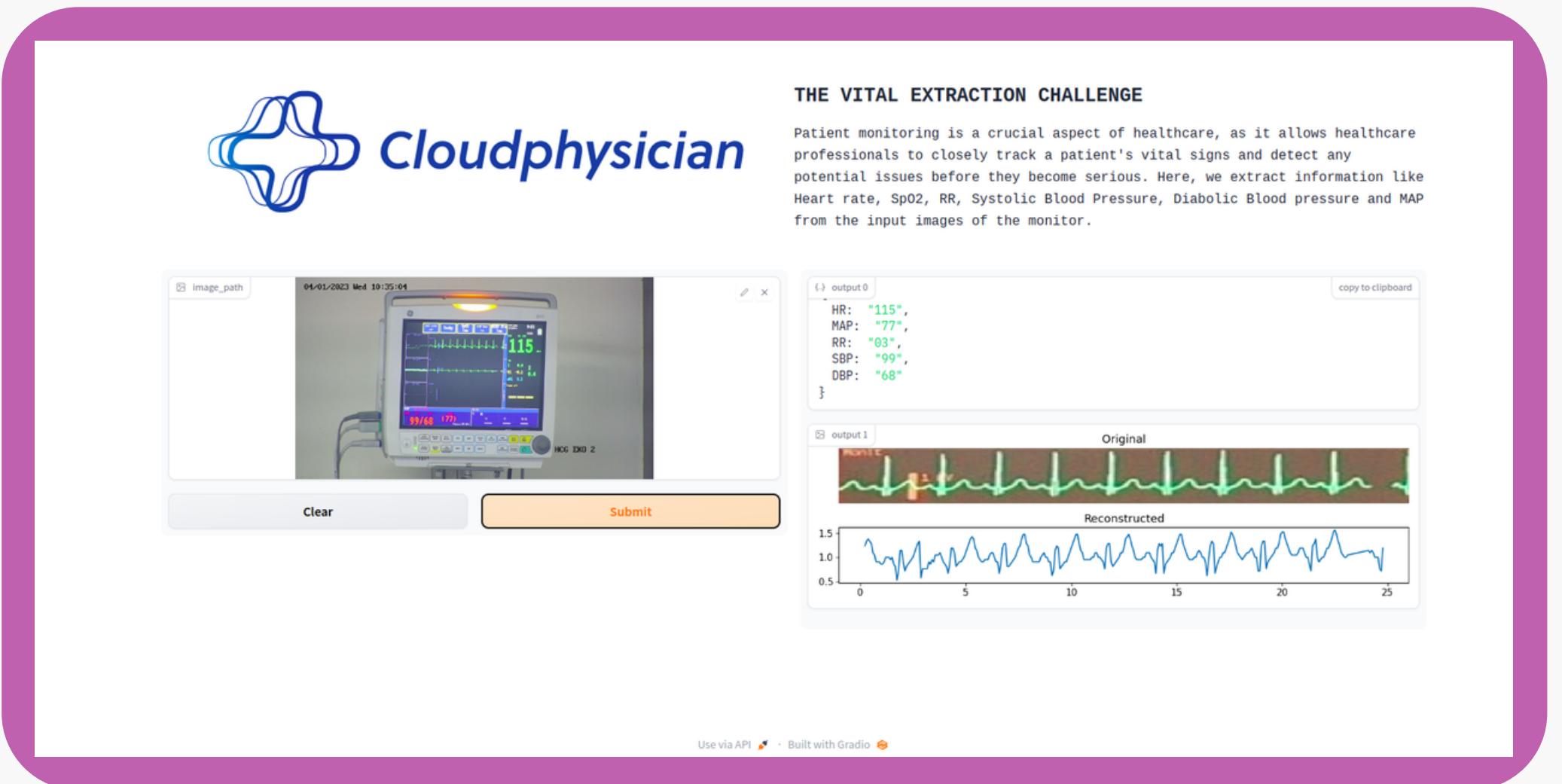
- **Post-processing** on the predicted results based on monitor layout info.
- **Segmentation instead of detection**
- **Semi-supervised** based segmentation approach to leverage unlabelled data.



Novelty

In terms of Visualization

- **Gradio visualisation** to show results - a library which runs in Colab itself.
- Provides a localhost to drop images and get results.



The screenshot shows a Gradio-based web application interface for extracting vital signs from patient monitoring images. The interface includes:

- A logo for "Cloudphysician" at the top left.
- A title "THE VITAL EXTRACTION CHALLENGE" at the top right.
- An input section labeled "image_path" containing a placeholder image of a medical monitor displaying vital signs.
- A "Clear" button and an orange "Submit" button below the input field.
- An "output 0" panel showing JSON extracted data:

```
{> output0
  HR: "115",
  MAP: "77",
  RR: "03",
  SBP: "99",
  DBP: "68"
}
```
- An "output 1" panel showing two心电图波形: "Original" (top) and "Reconstructed" (bottom), along with a line graph plot.
- A "copy to clipboard" button in the top right corner of the output panels.
- Small links at the bottom: "Use via API" and "Built with Gradio".

Thank You