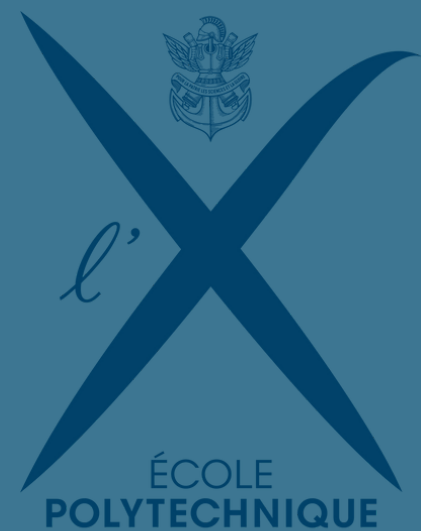


SEGMENTATION OF CANCER CELLS IN 3D LUNGS CT SCANS

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INF573 : Computer Vision Defense

Computed Tomography (CT) Scans

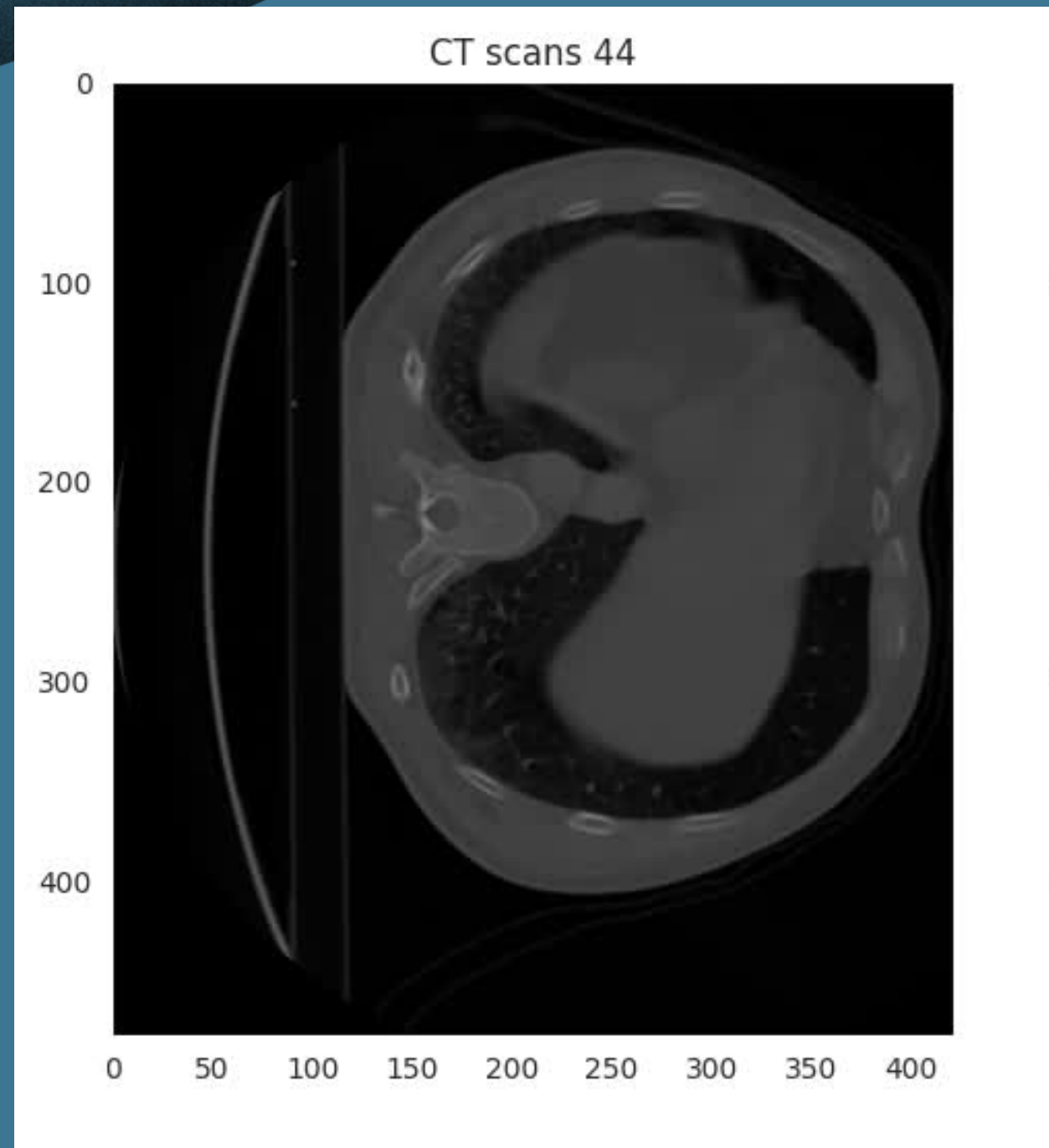


Figure 1 : Example of a CT Scan



Figure 2 : Scanner

Computed Tomography (CT) Scans

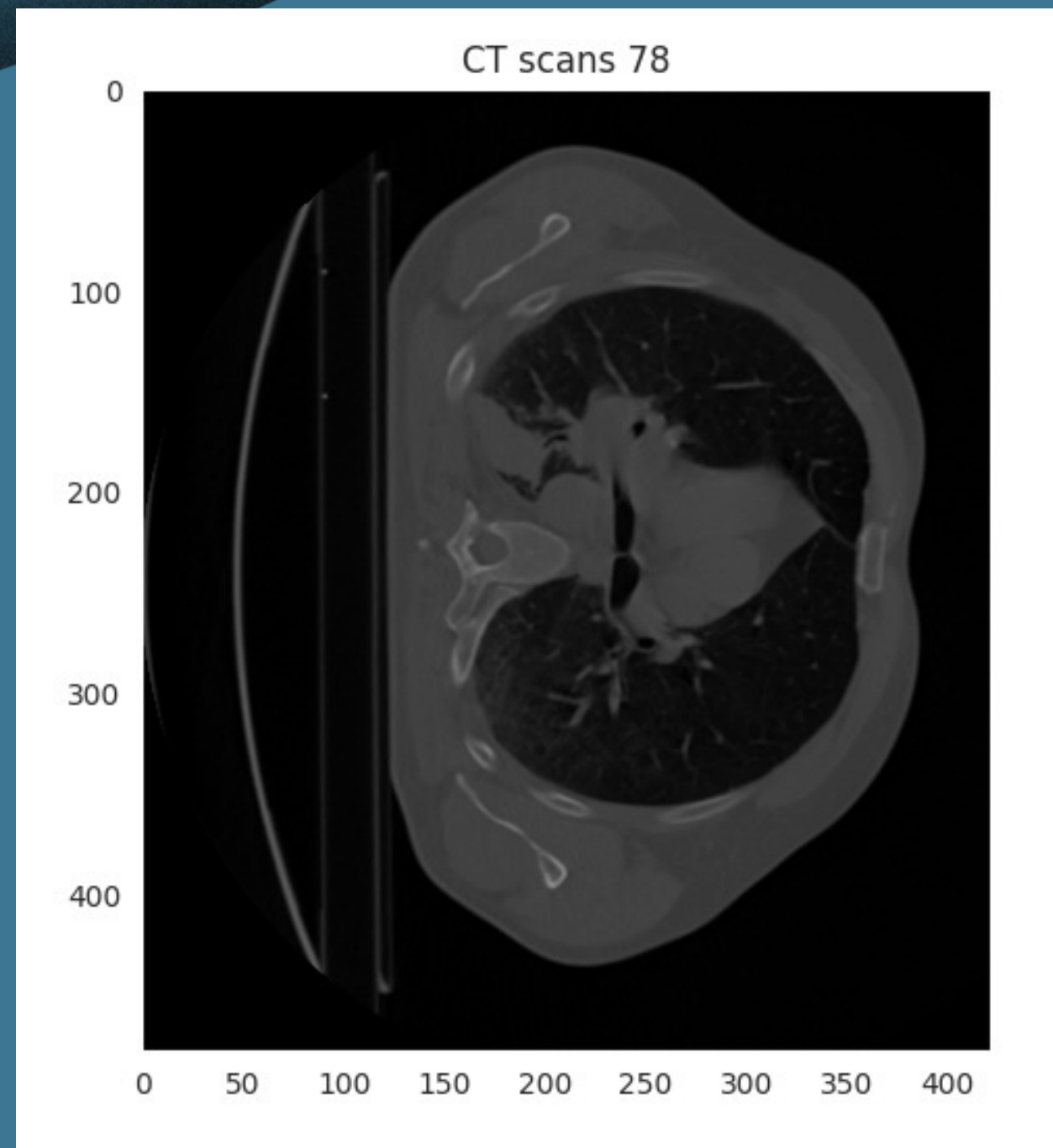


Figure 1 : Example of a CT Scan

- Shape : $512 \times 512 \times N$ where N is around 110
- Size : around 1mm x 1mm x 3 mm
- Training
 - 291 Files
 - Scans, Segmentation
- Test
 - 100 Files
- Task : 3D Segmentation of Cancer Cells

Computed Tomography (CT) Scans

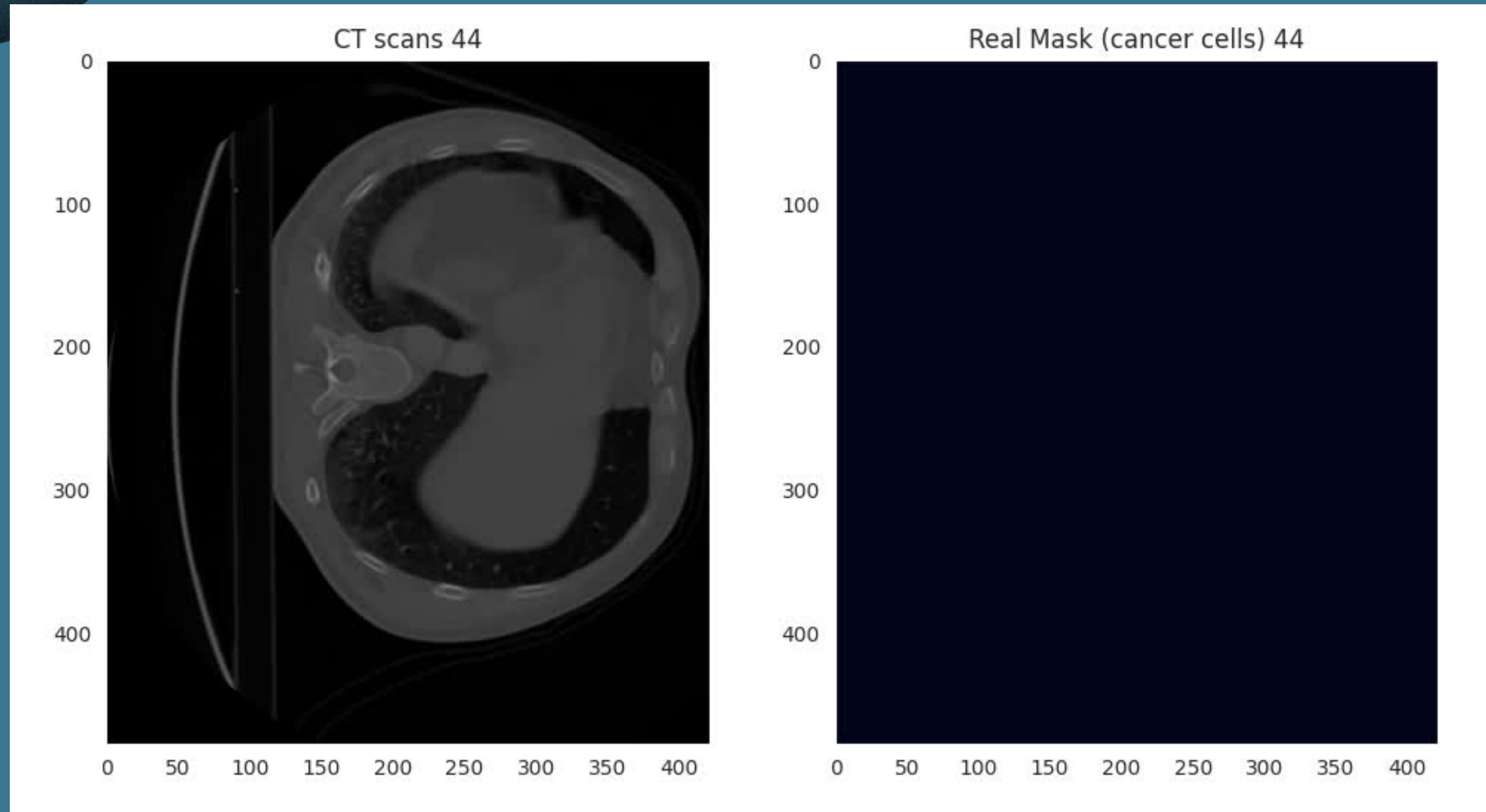


Figure 3 : CT Scan (left), Binary Mask for main tumor (right)

Measuring performance : Dice Metric

$$\text{Dice Loss} = 1 - \frac{2|A \cap B|}{|A| + |B|}$$

$$\text{Dice Metric} = \frac{2|A \cap B|}{|A| + |B|}$$

Figure 4 : Dice Metric

Note : it corresponds to Dice Metric corresponds to F1

Architecture : 3D Unet

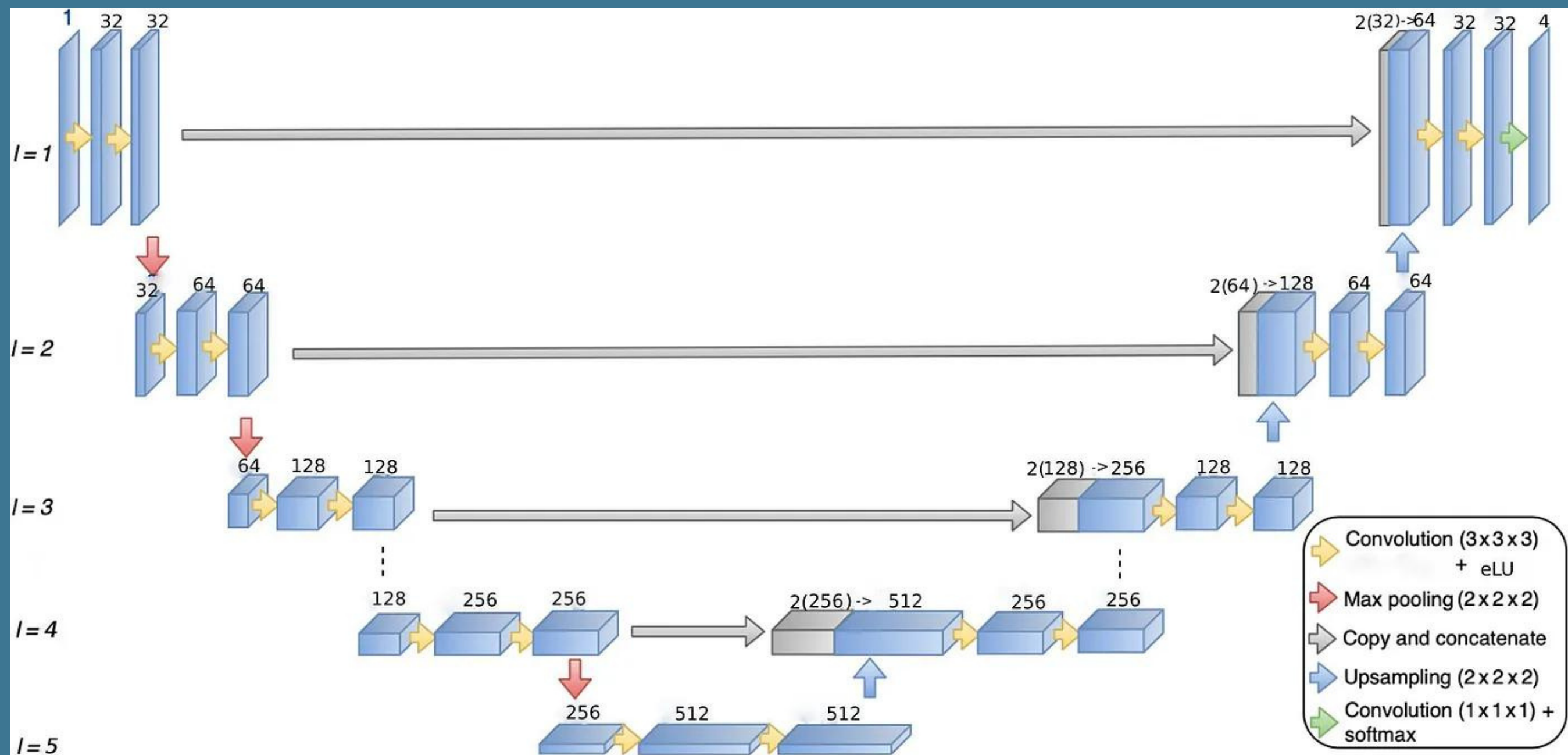


Figure 5 : Architecture of a 3D Unet

Data - Pipeline

Preprocessing

- Crop Image (remove border)
- Ensure orientation
- Scale Intensity to $[0, 1]$ (from $[-1024, 3071]$)
- Random extract patch (center selected using its label)

Postprocessing

- Softmax
- Keep Largest Connected Component
- Fill in

Prediction - Sliding Window

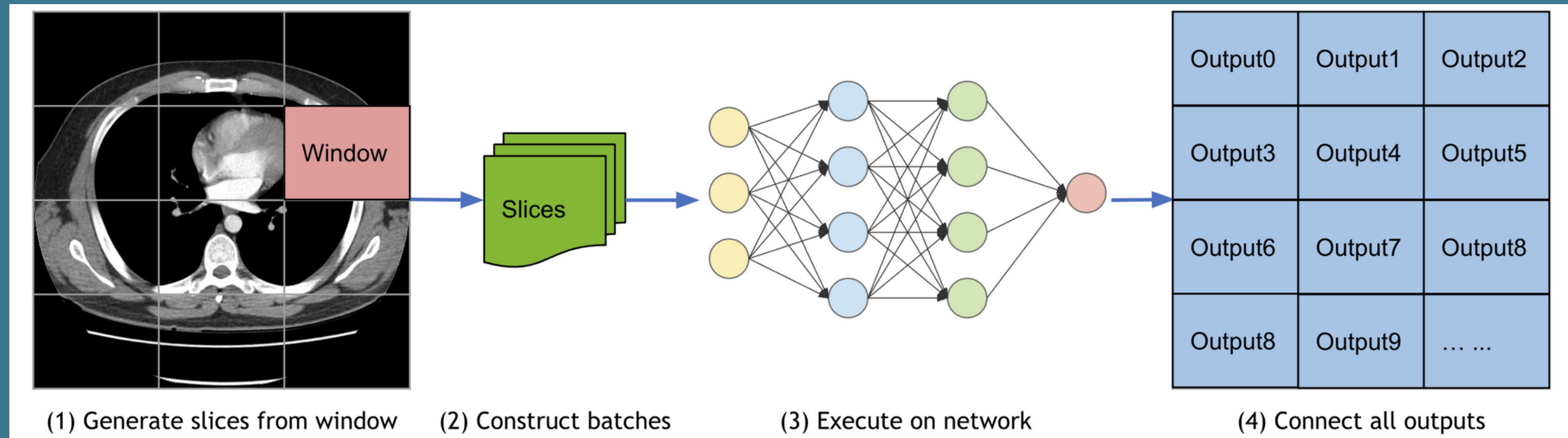


Figure 6 : Explanation Sliding Window Inference

Training

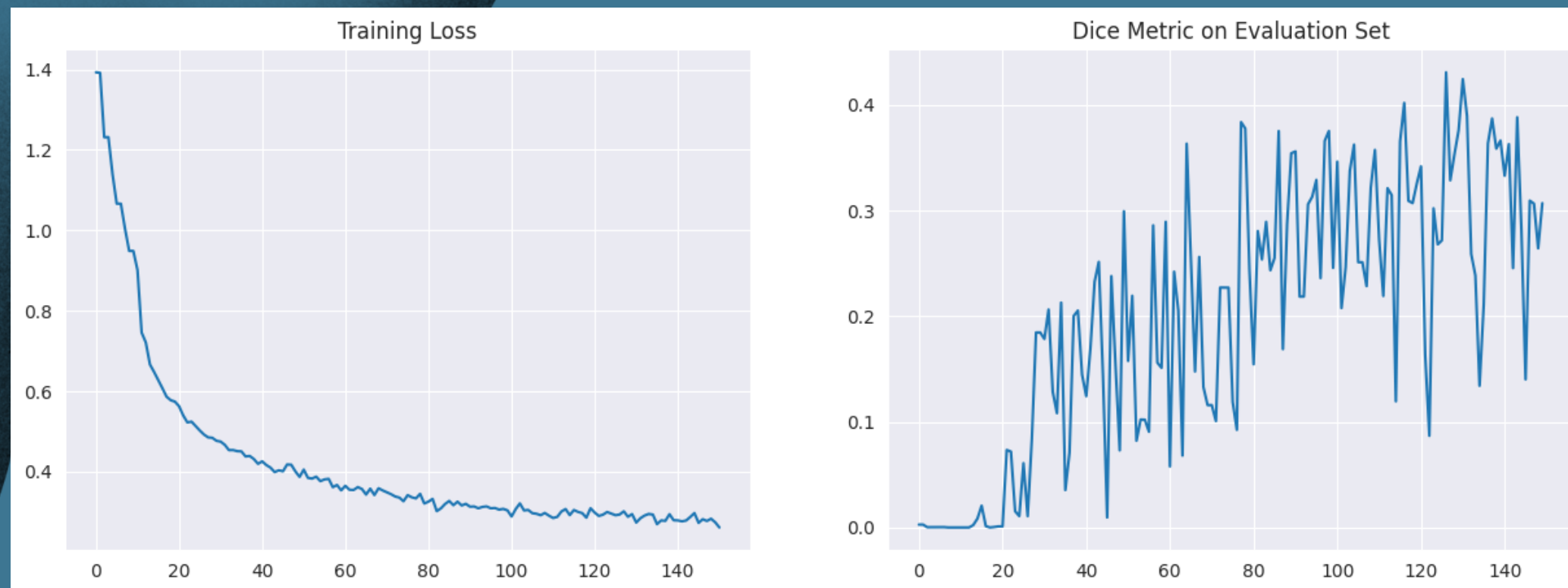


Figure 7 : Unet 150 epochs

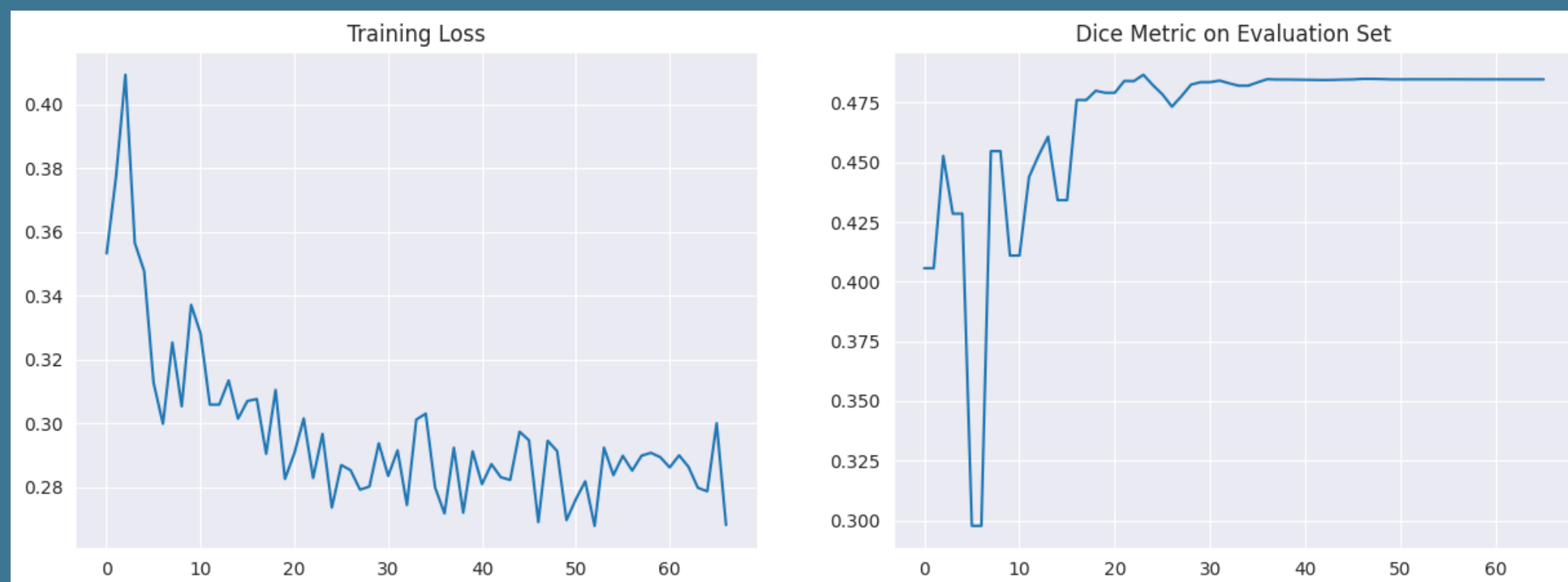


Figure 8 : Further training of previous model

Note :

- Better postprocessing
- Increase exposure to negative sampling
- Too strong reduction of learning rate

Results

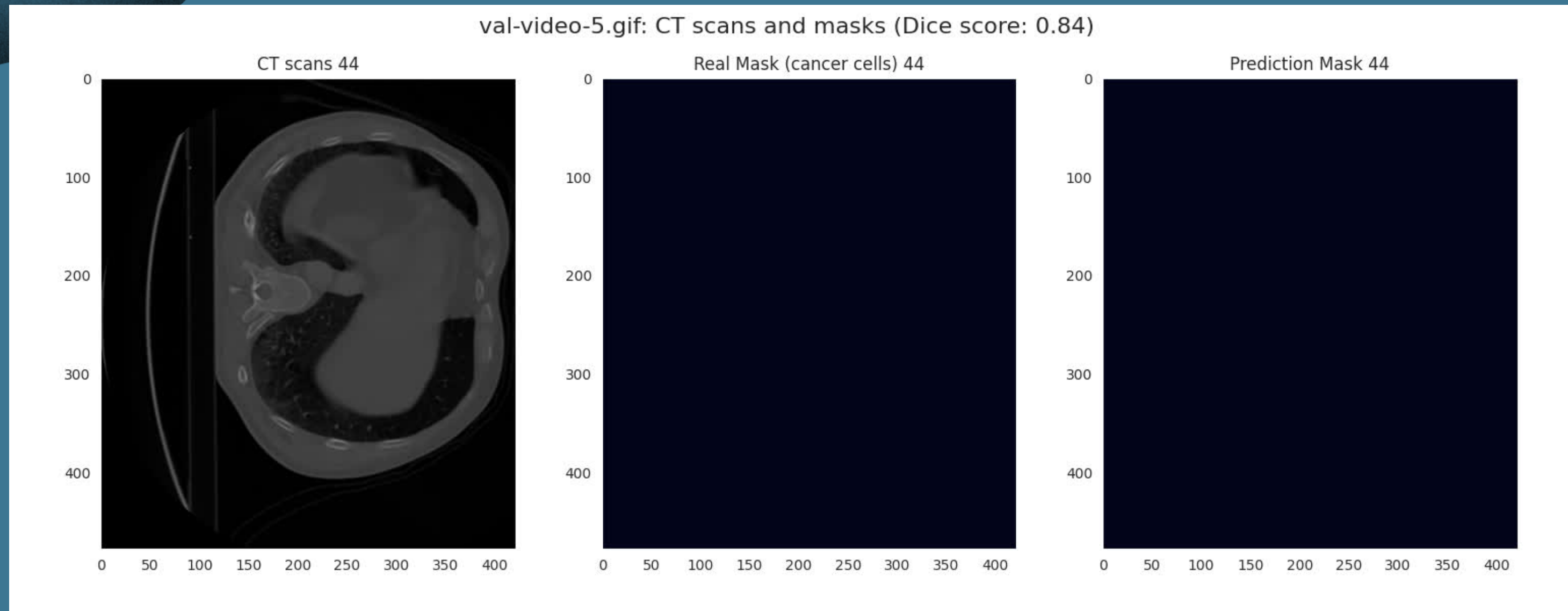


Figure 9 : Prediction on evaluation sample