



MEDICAL IMAGE SEGMENTATION USING CNN AND DYNAMIC PROGRAMMING



Nilabjanayan Bera

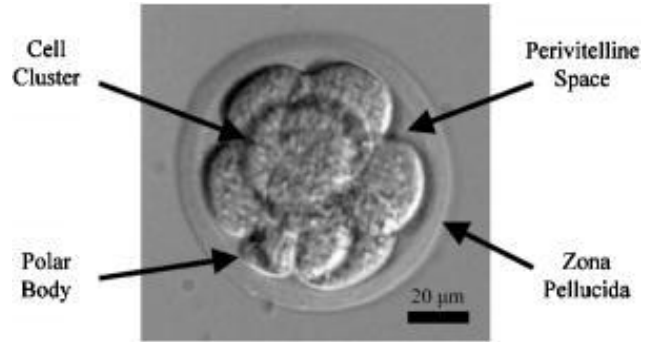




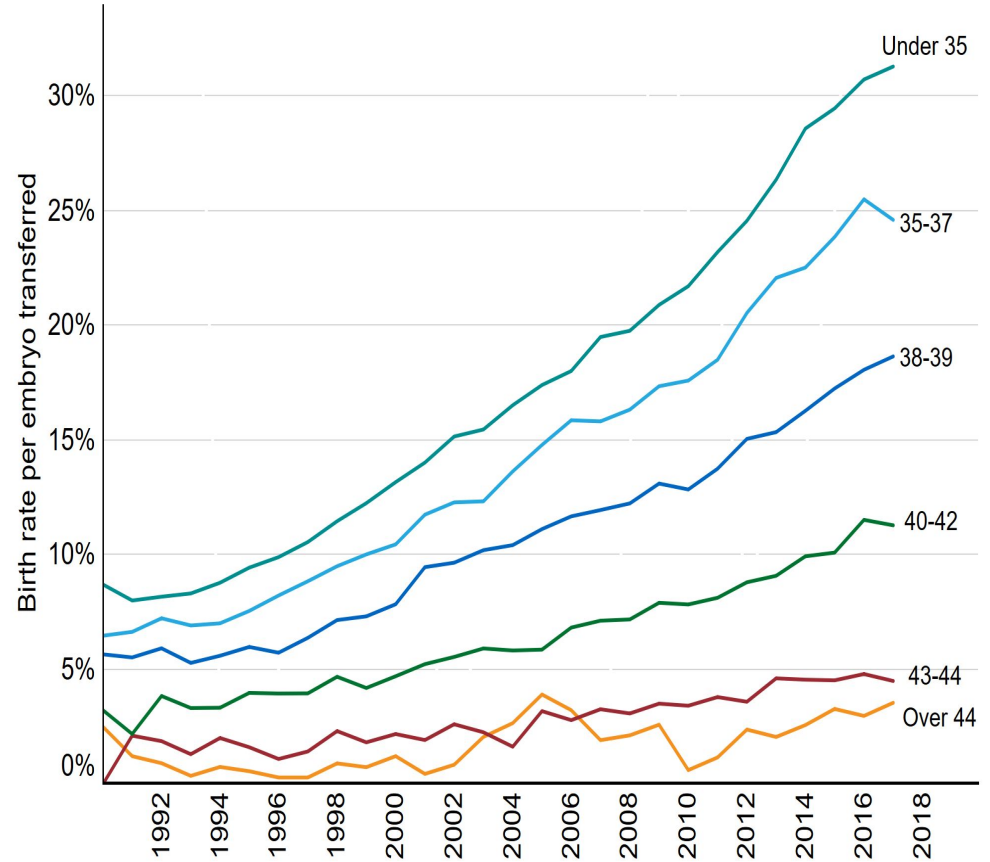
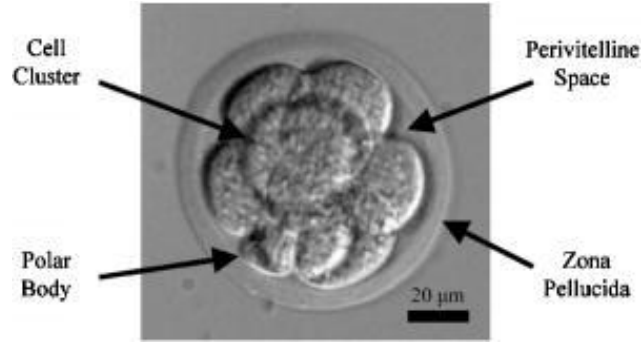
Medical Segmentation Tasks : Lack of Training Data



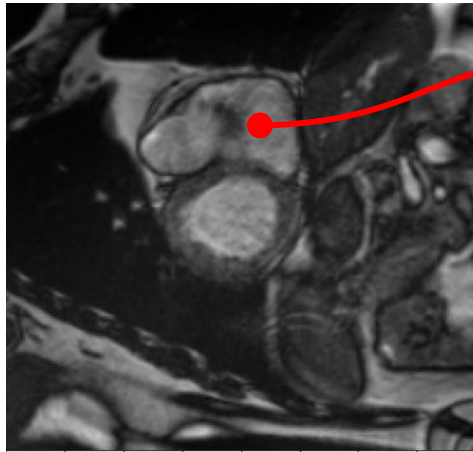
CASE - I



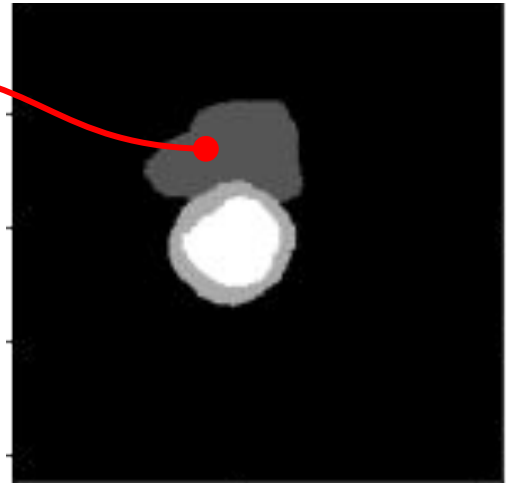
CASE - I



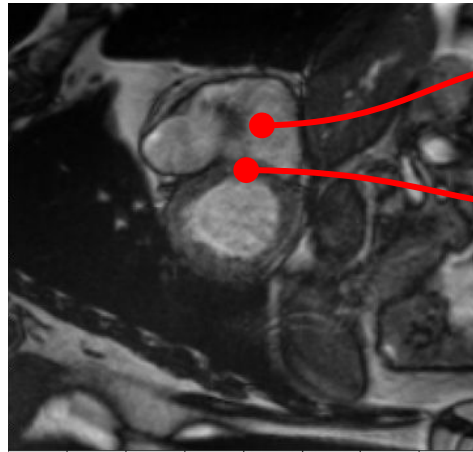
CASE - II



Right Ventricle

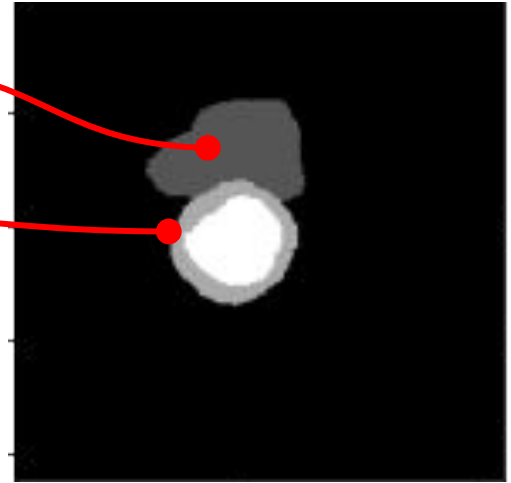


CASE - II

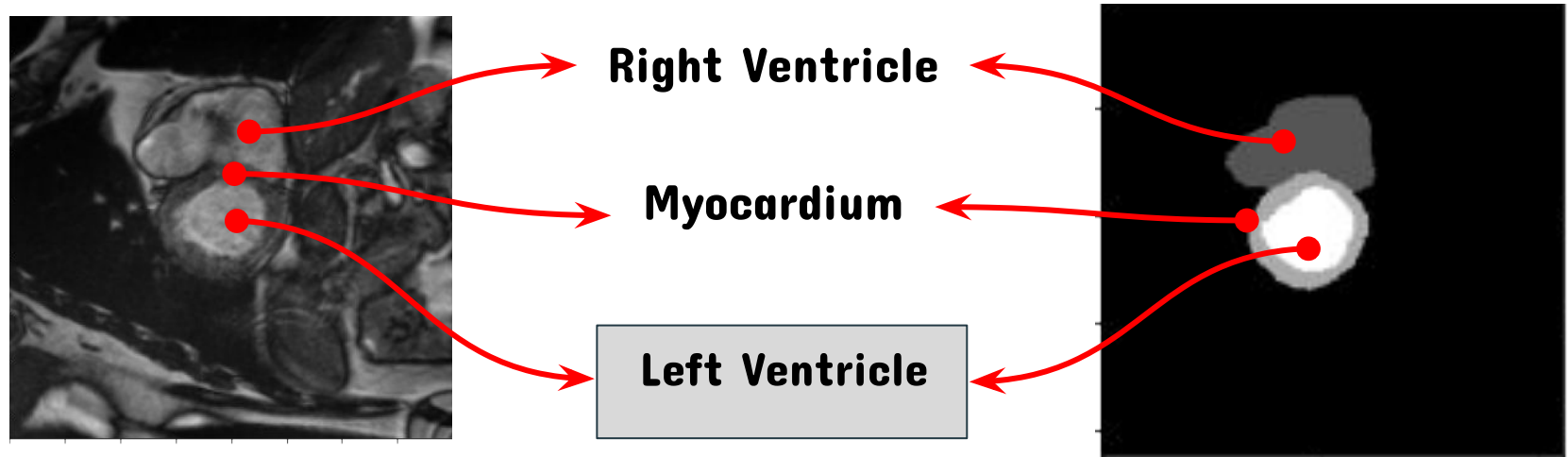


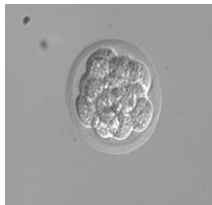
Right Ventricle

Myocardium



CASE - II





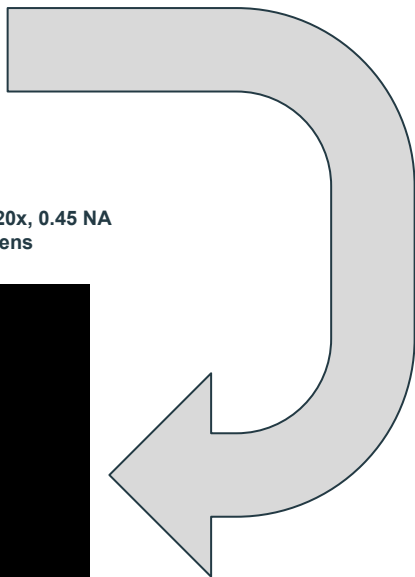
(640, 480)

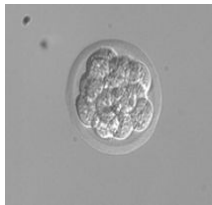
Nikon Eclipse TE200 microscope with a 20x, 0.45 NA objective lens and a 0.52 NA condenser lens



Training : 12

Testing : 3





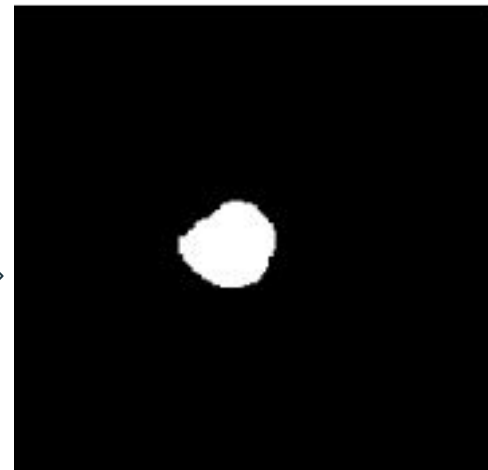
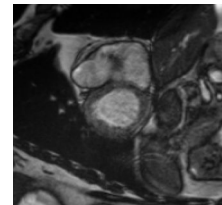
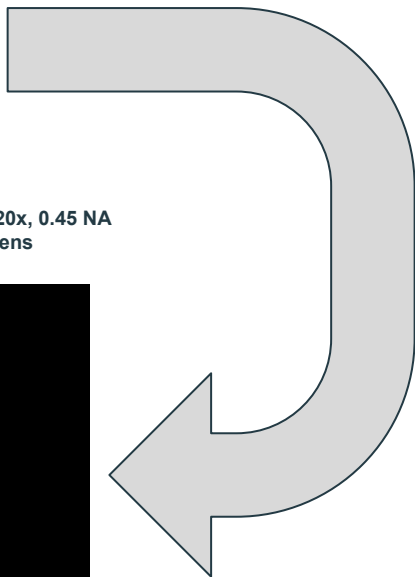
(640, 480)

Nikon Eclipse TE200 microscope with a 20x, 0.45 NA objective lens and a 0.52 NA condenser lens



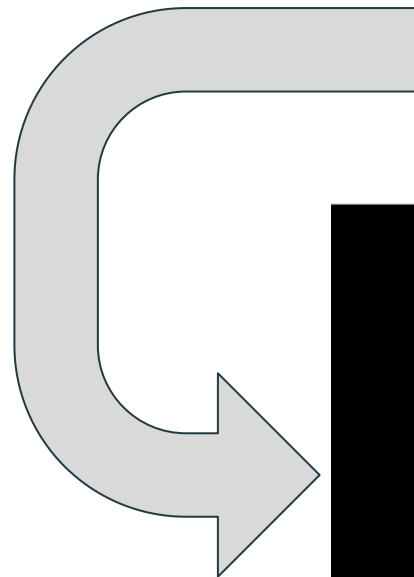
Training : 12

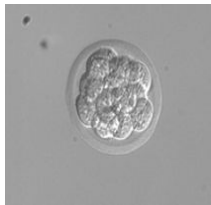
Testing : 3



Training : 1516

Testing : 386





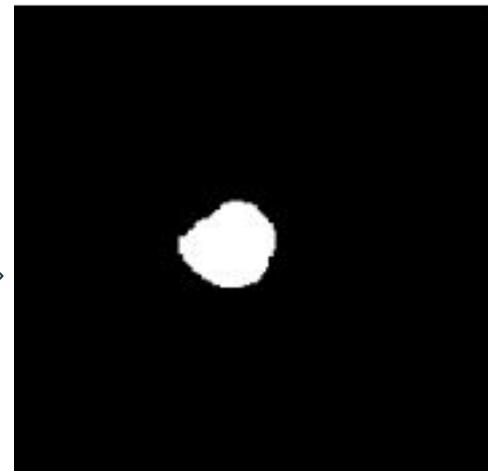
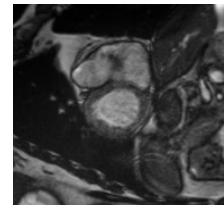
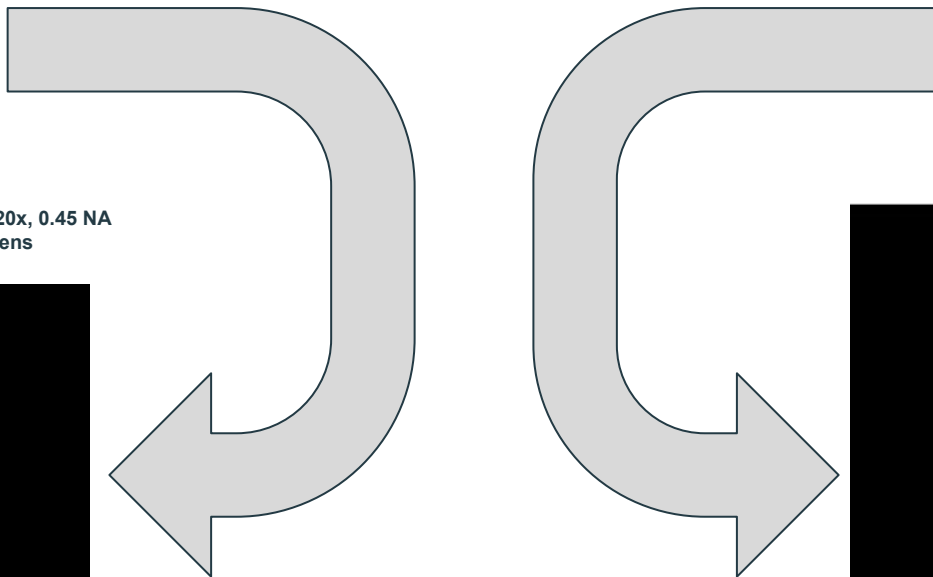
(640, 480)

Nikon Eclipse TE200 microscope with a 20x, 0.45 NA objective lens and a 0.52 NA condenser lens



Training : 12

Testing : 3



Training : 1516

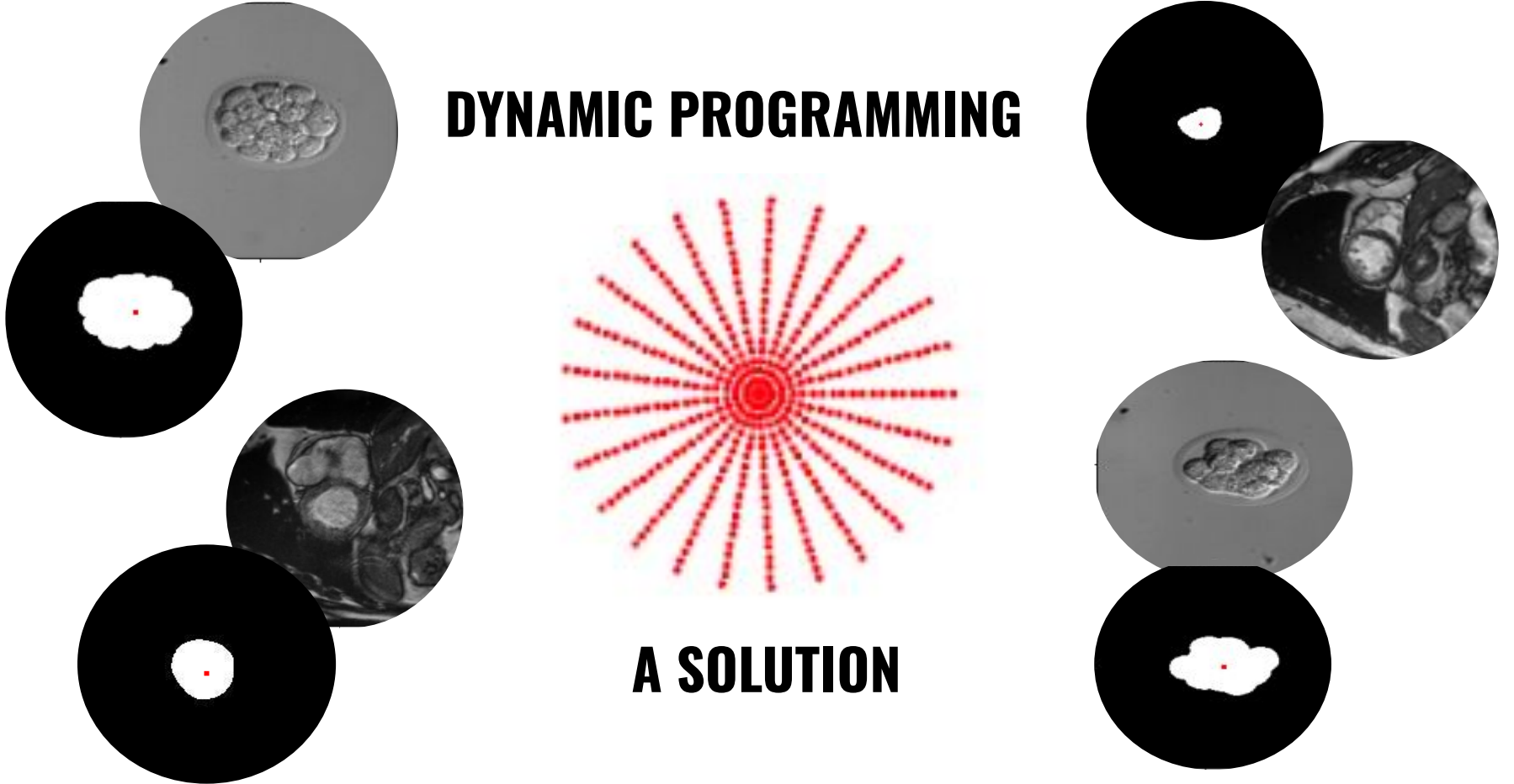
Testing : 386

**Small no. of
training data!!**

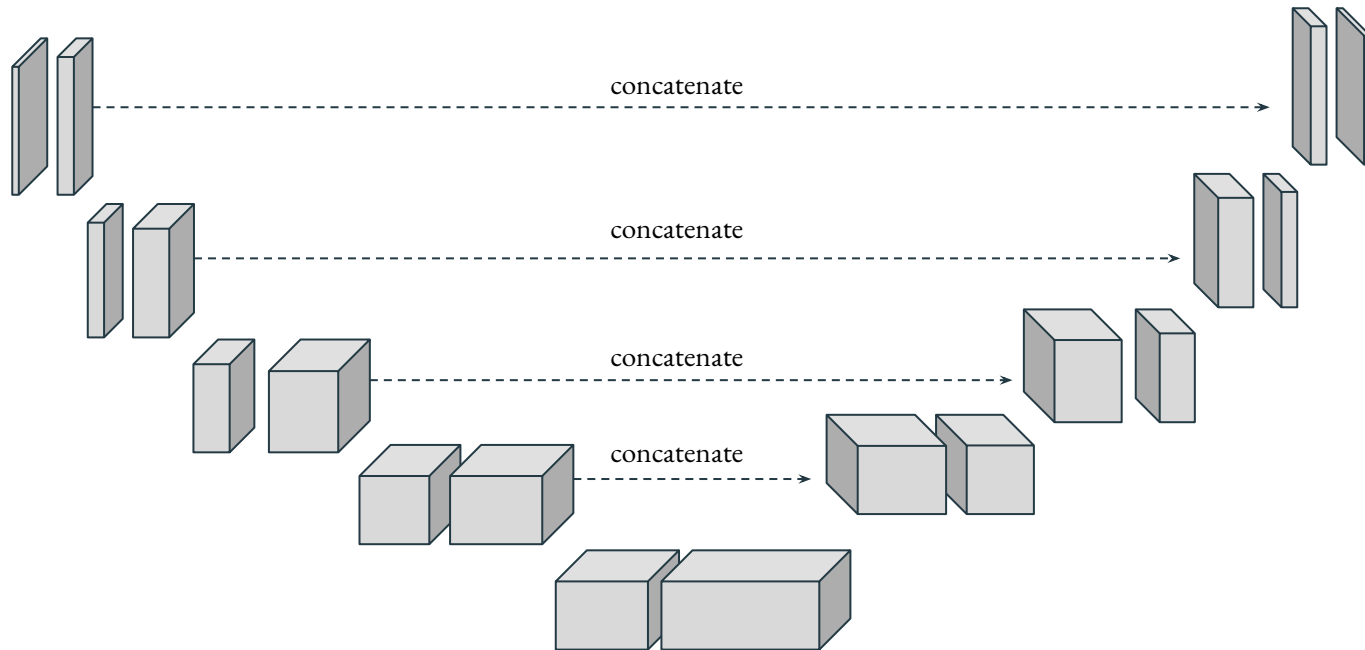
DYNAMIC PROGRAMMING



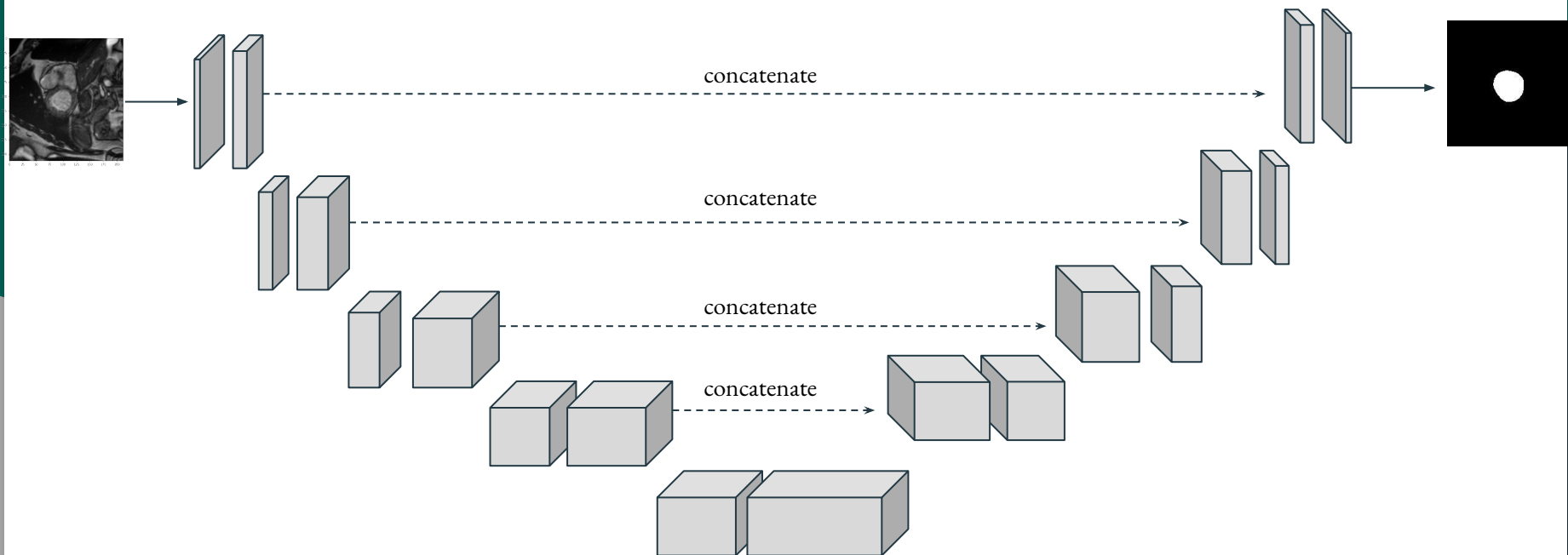
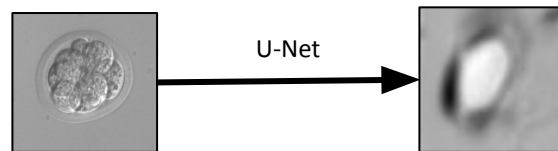
A SOLUTION

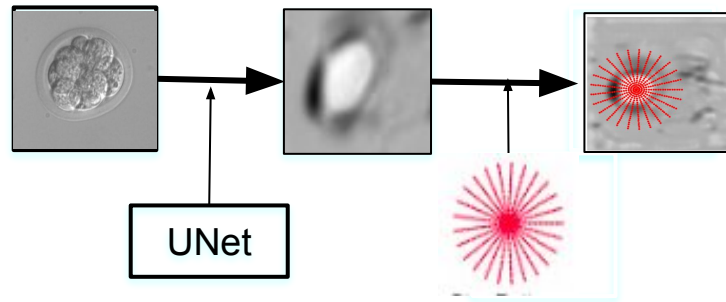


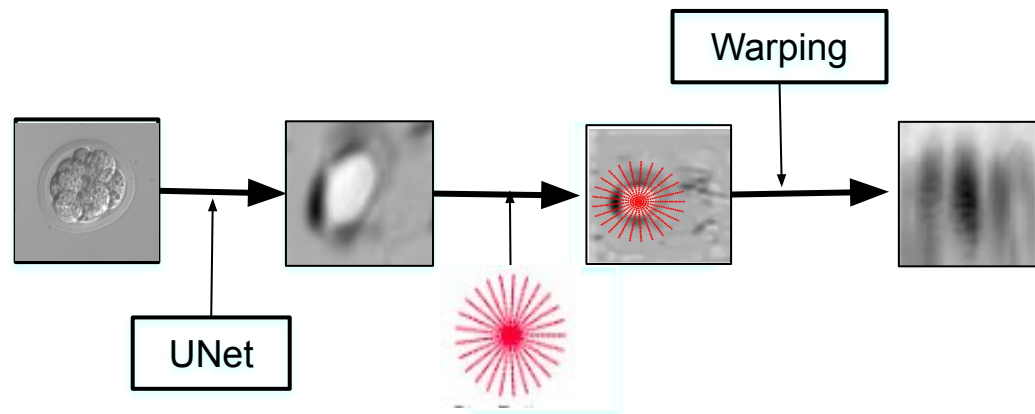
U-Net

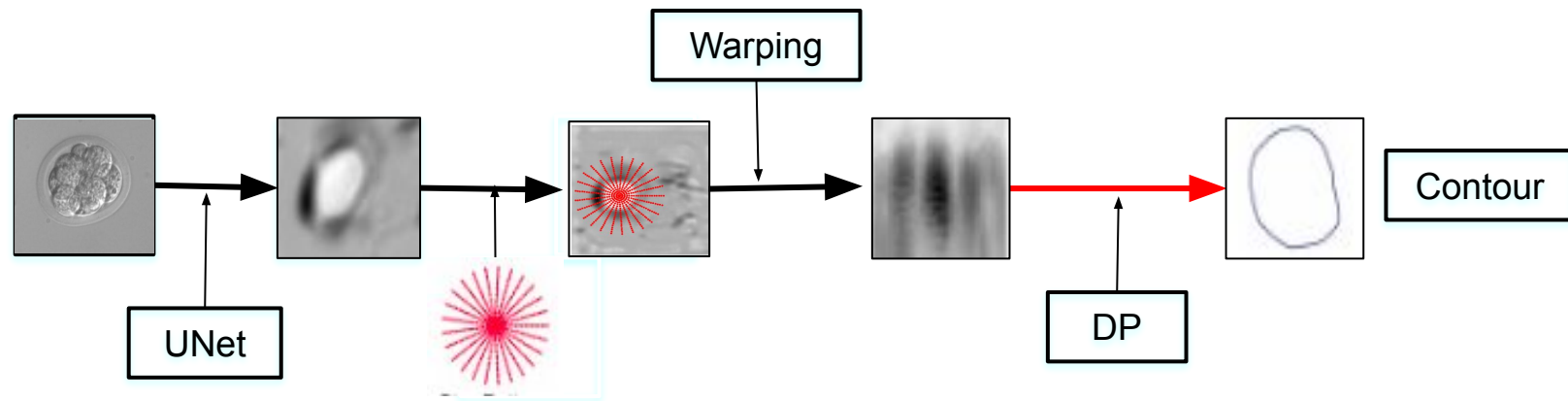


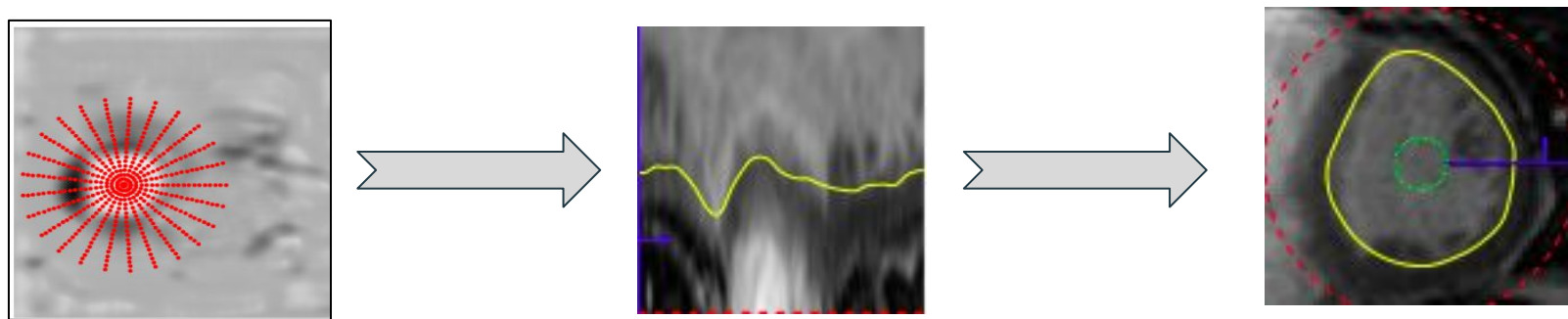
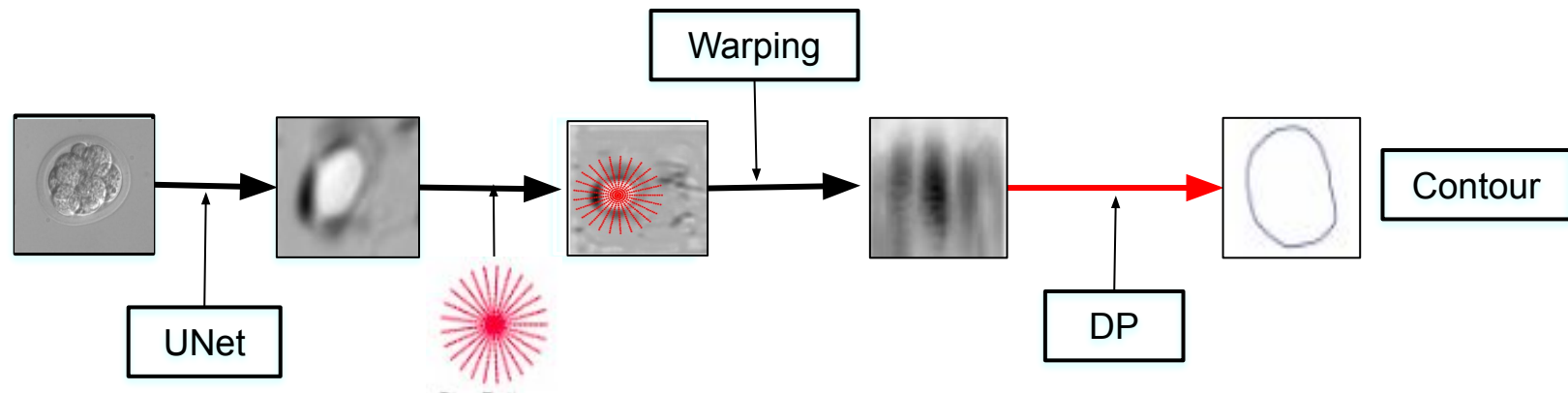
U-Net











Dynamic Programming



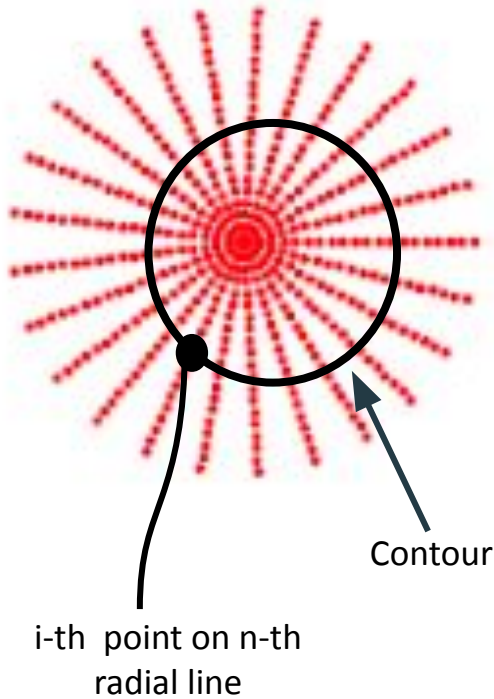
No. of radial lines : N

No. of candidate points on each line : M

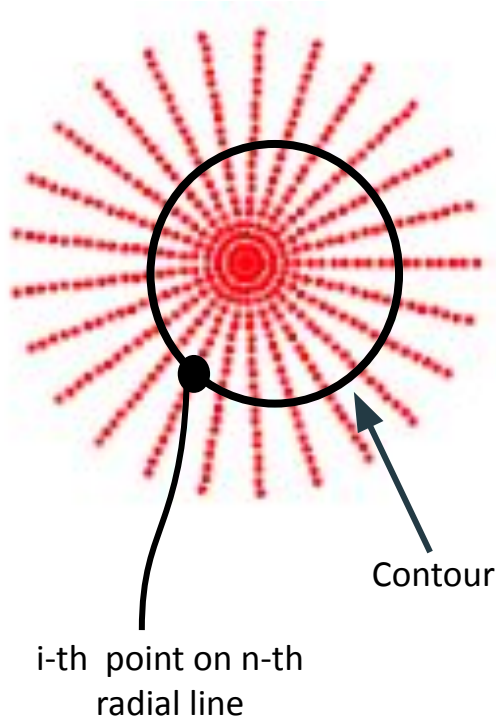
Dynamic Programming

No. of radial lines : N

No. of candidate points on each line : M



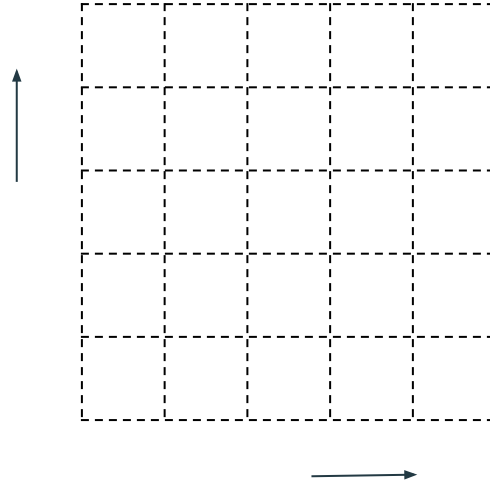
Dynamic Programming



No. of radial lines : N

No. of candidate points on each line : M

$g(n,i)$: value of warped map on the i -th point on the n -th radial line.



Dynamic Programming



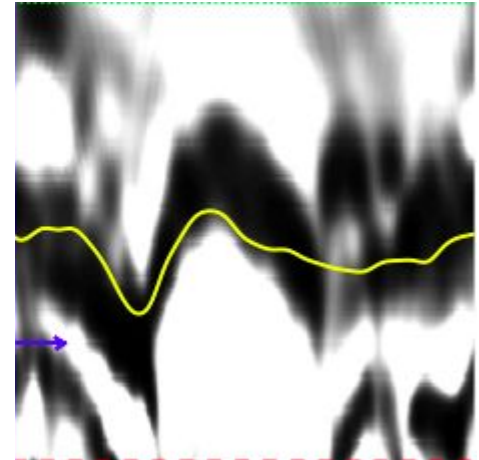
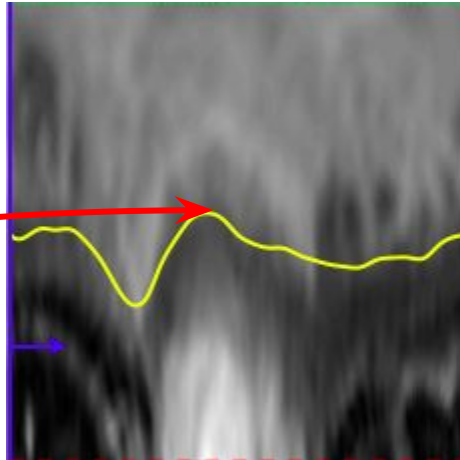
i-th point on n-th
radial line

No. of radial lines : N

No. of candidate points on each line : M

$g(n, i)$: value of warped map on the i-th point on the n-th radial line.

$dg(n, i) = g(n, i) - g(n, i-1)$, directional derivative on g



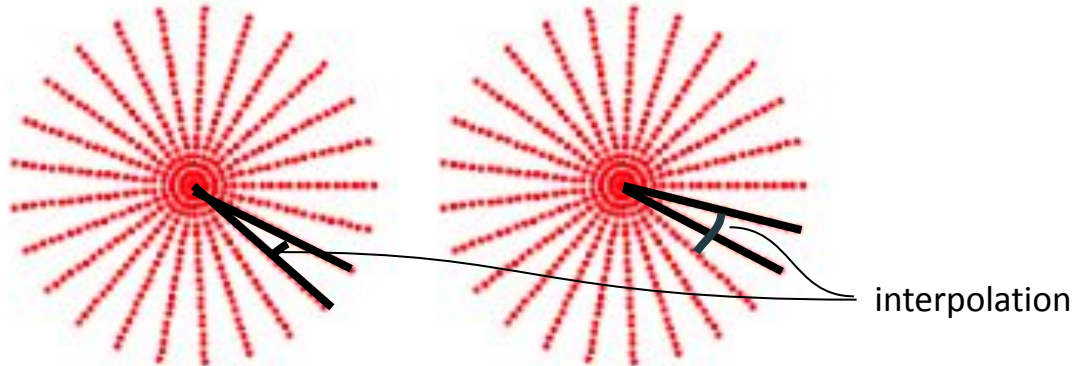
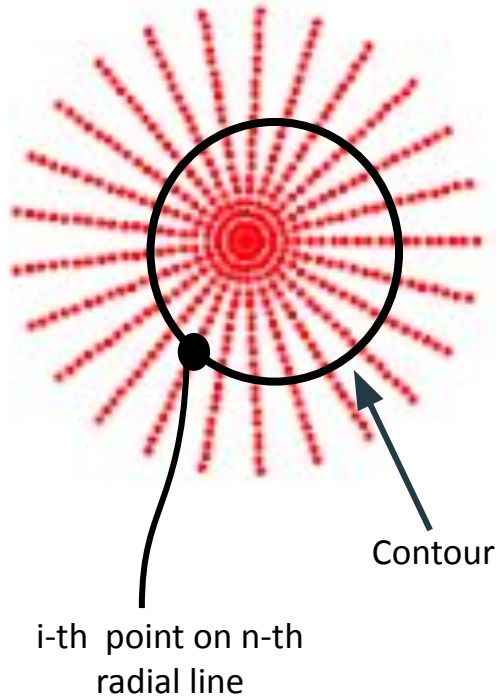
Dynamic Programming

$dg(n, i) = g(n, i) - g(n, i-1)$, directional derivative on g

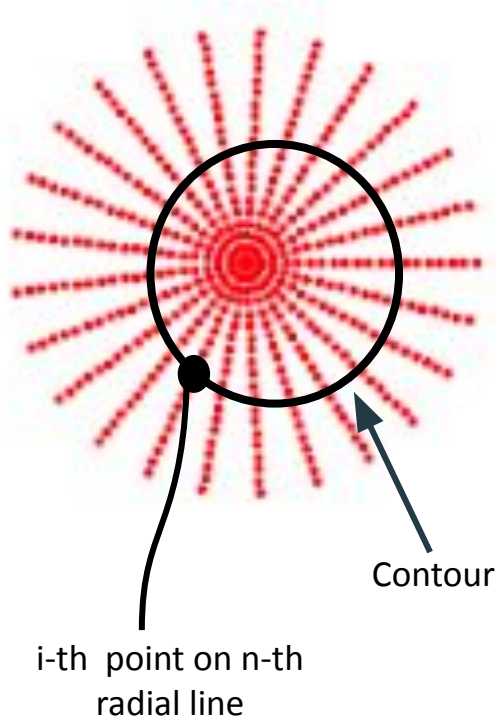
$$E(n, i, j) = \begin{cases} dg(n, i) + dg(n \oplus 1, j), & \text{if } |i - j| \leq \delta \\ \infty, & \text{otherwise,} \end{cases}$$

Our cost function:

$$\min_{v_1, \dots, v_N} E(N, v_N, v_1) + \sum_{n=1}^{N-1} E(n, v_n, v_{n+1})$$



Dynamic Programming



No. of radial lines : N

No. of candidate points on each line : M

$g(n, i)$: value of warped map on the i -th point on the n -th radial line.

$dg(n, i) = g(n, i) - g(n, i-1)$, directional derivative on g

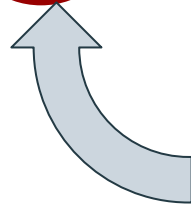
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Our cost function:

$$\min_{v_1, \dots, v_N} E(N, v_N, v_1) + \sum_{n=1}^{N-1} E(n, v_n, v_{n+1})$$

Loss function:

$$\min_{v_1, \dots, v_N} E(N, v_N, v_1) + \sum_{n=1}^{N-1} E(n, v_n, v_{n+1})$$



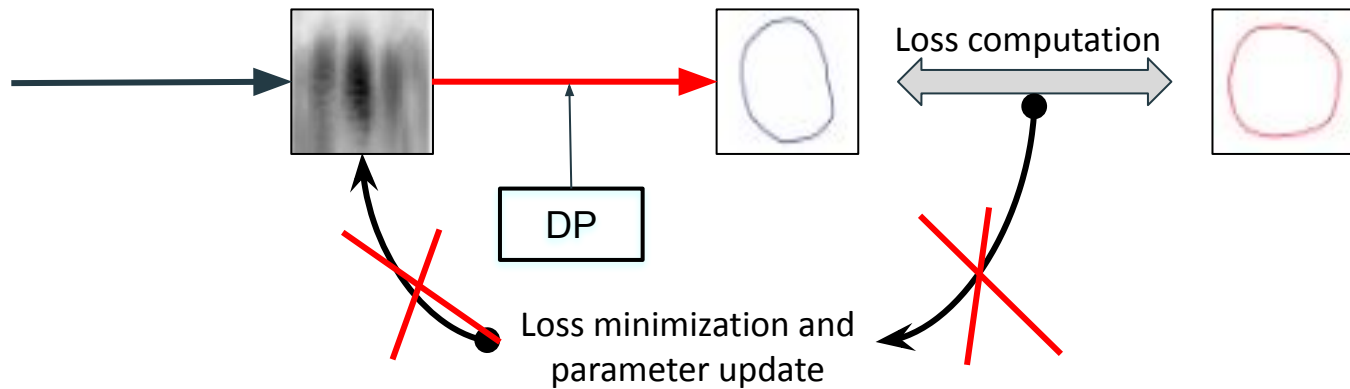
argmin !!

Differentiable??

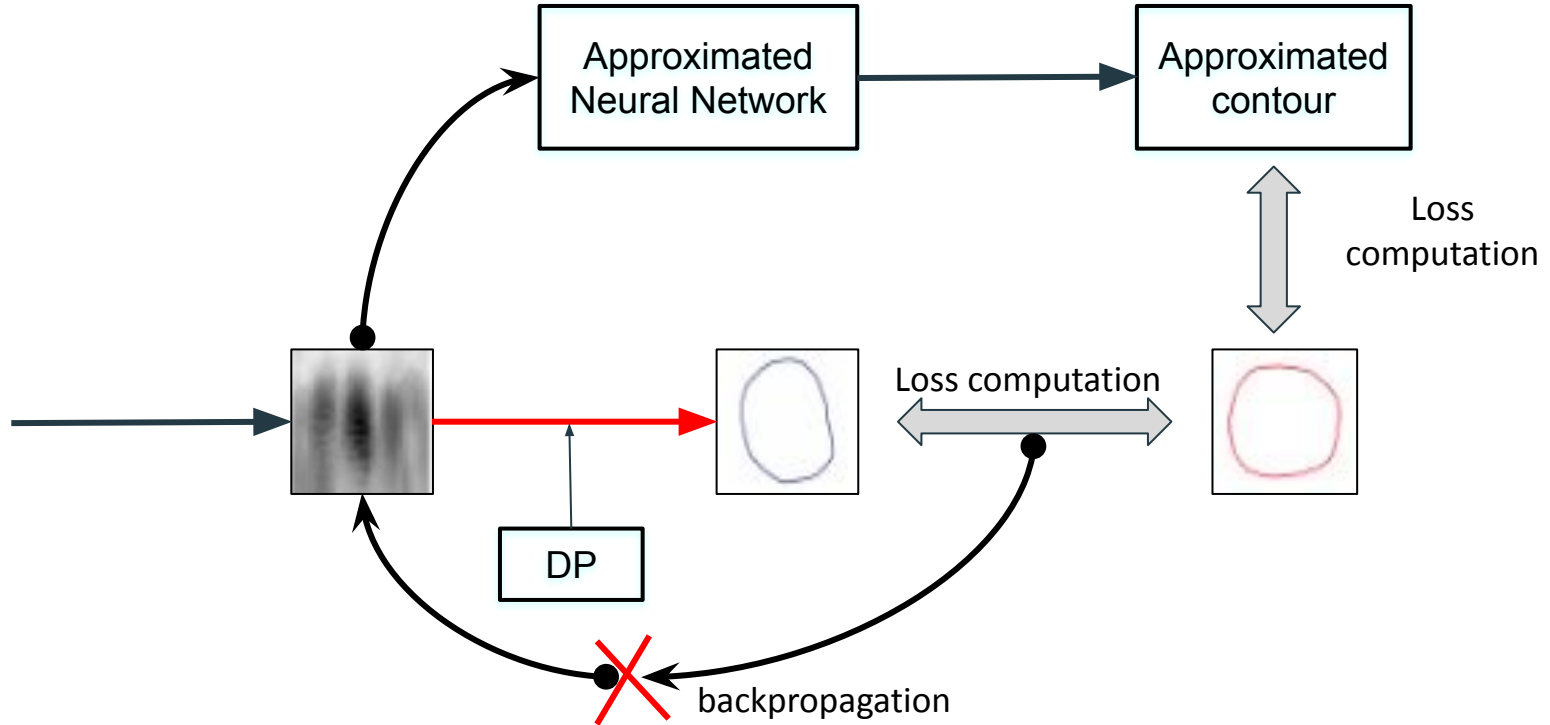
Loss function:

$$\min_{v_1, \dots, v_N} E(N, v_N, v_1) + \sum_{n=1}^{N-1} E(n, v_n, v_{n+1})$$

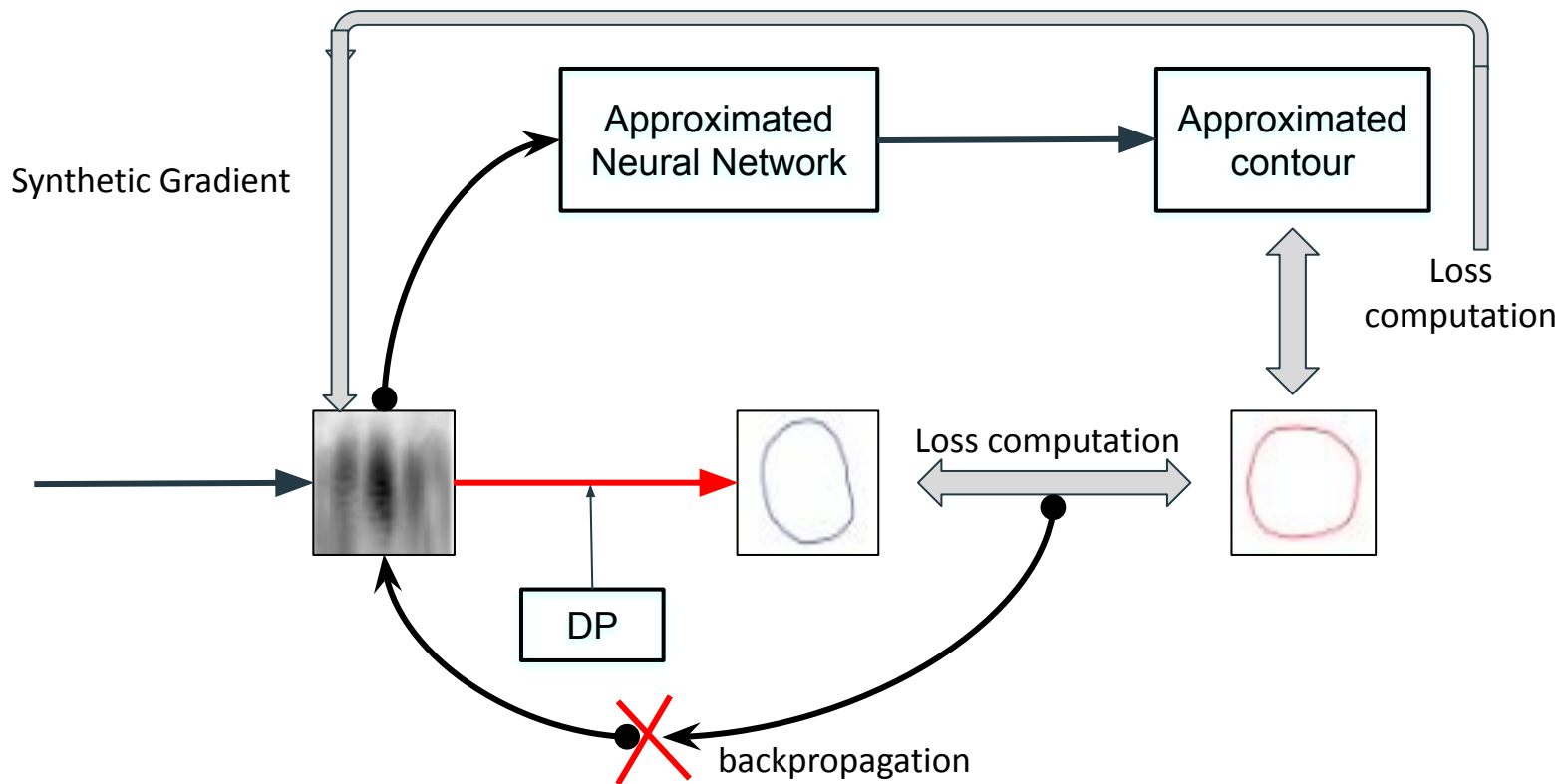
argmin !!

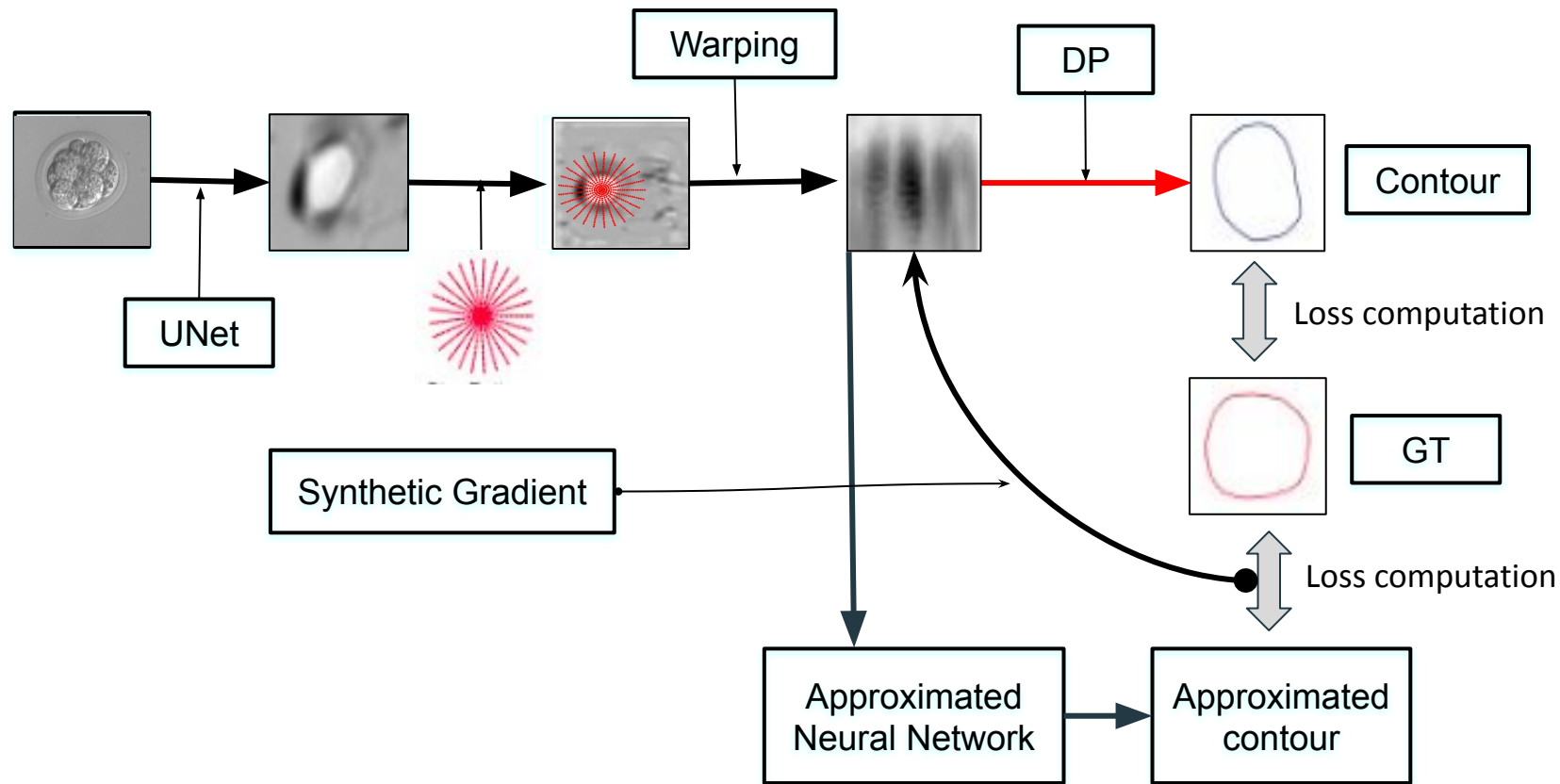


Approximated Neural Network



Approximated Neural Network





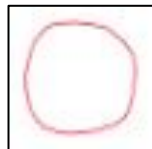
Introduce the randomness



$DP(g)$



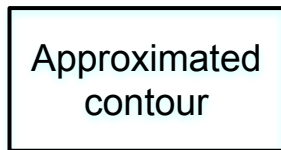
Exact loss



Ground Truth



Mimicked loss



$F(g)$

For the DP module, the output is $DP(g)$

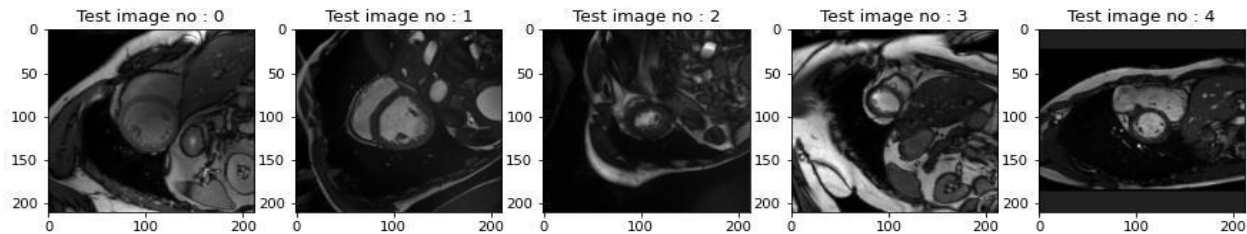
For the approximated module, the output is $F(g)$

Mimicked loss: $L(F(g), DP(g))$

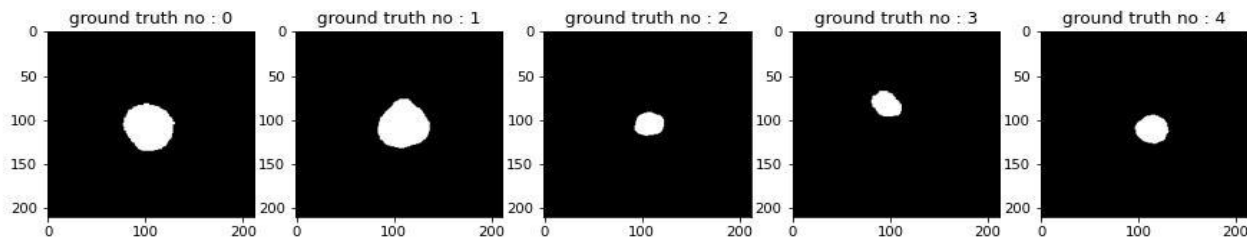
Randomized loss: $L(F(g + \sigma\epsilon), DP(g + \sigma\epsilon))$

RESULTS

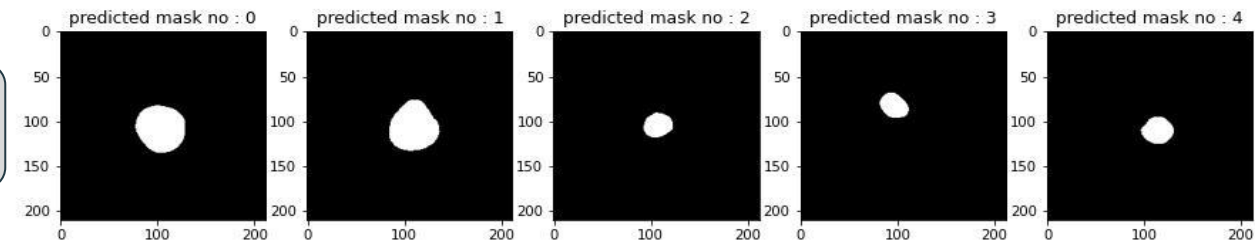
Images



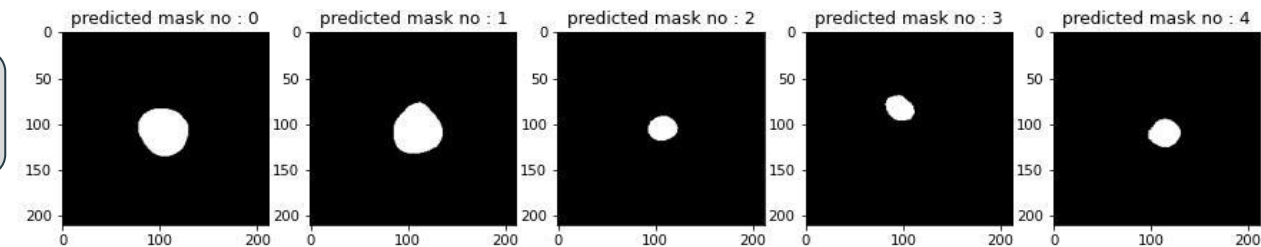
Ground Truth



U-Net



EDPCNN

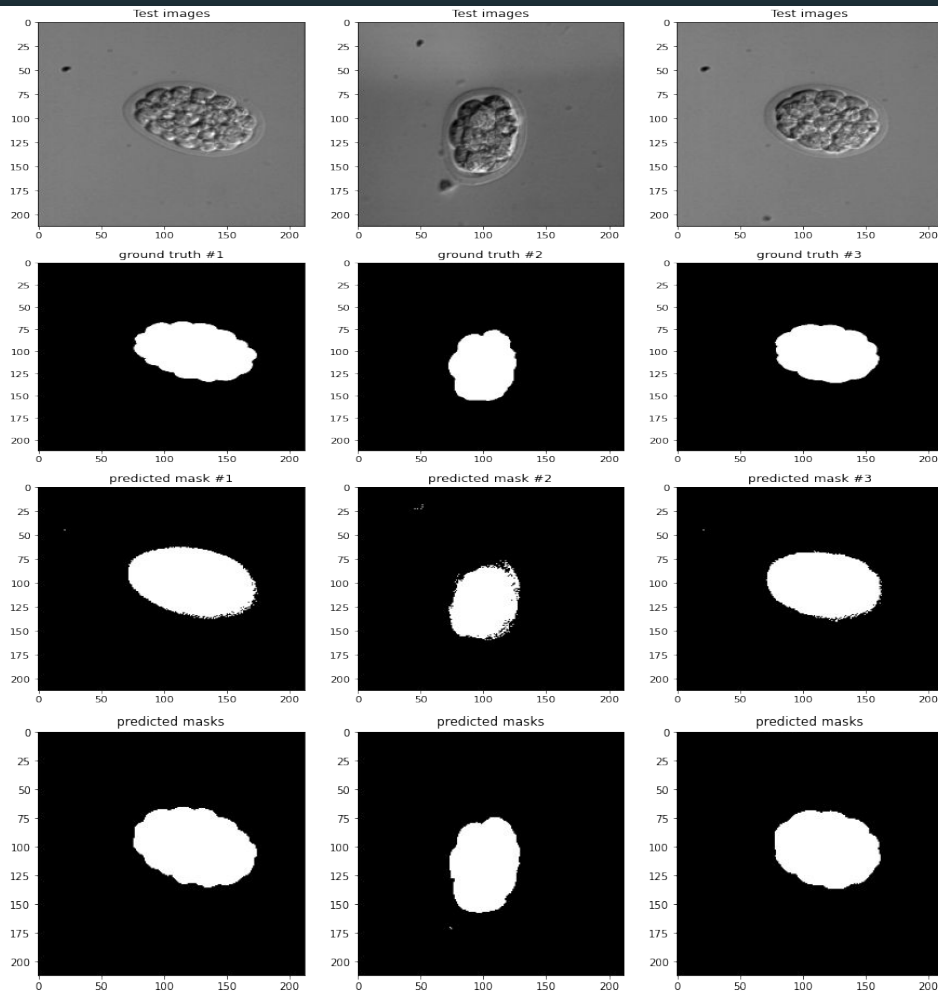


Images

Ground Truth

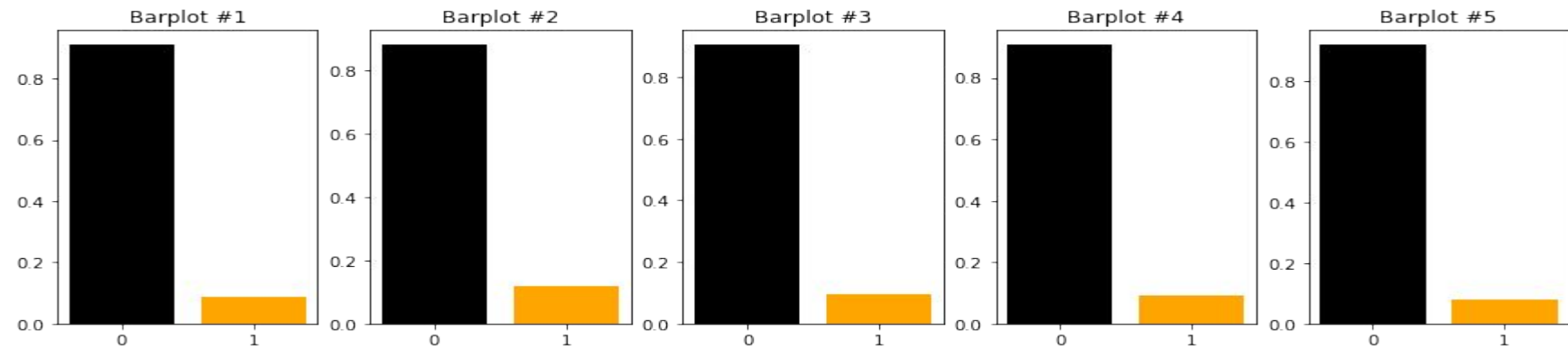
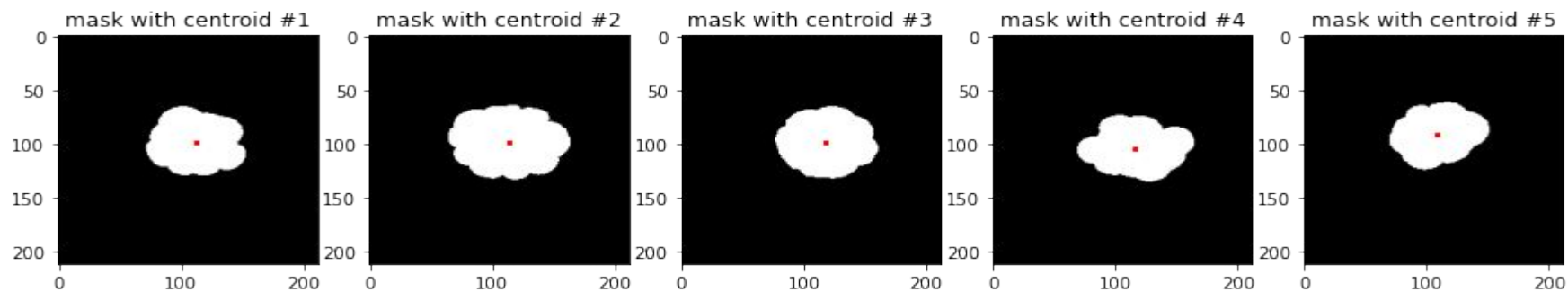
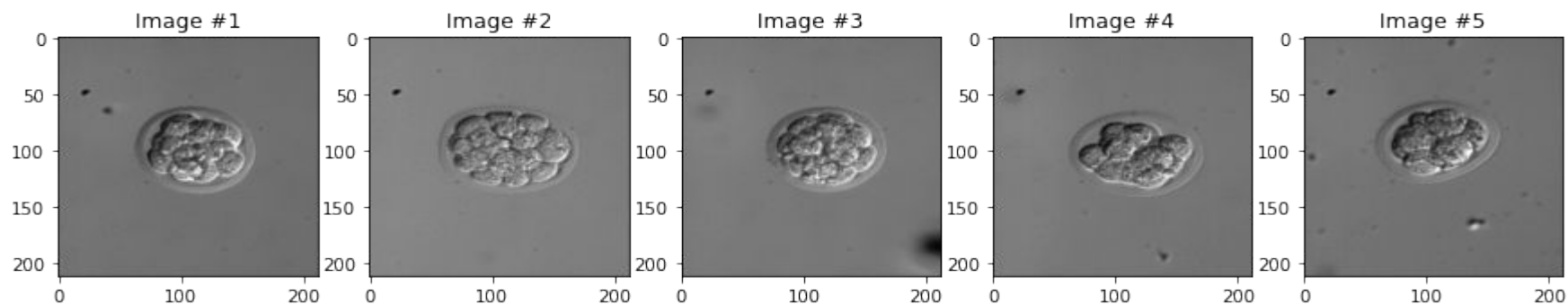
U-Net

EDPCNN





CONCLUSION



Further Scope of Improvements

- Cell no. detection in embryo.
- Segment myocardium and right ventricle or multiple cells with automated placement of multiple star patterns.
- Disease prediction from heart masks.



THANK YOU