

International
Institute of Information
Technology Bangalore

Hardware-software co-design to accelerate Brain Tumour Segmentation

VINAY RAYAPATI

GOPALA KRISHNA REDDY SANAMPUDI

AJAY KUMAR GANDI

RAVI KIRAN REDDY GOGIREDDY

Prof. NANDITHA RAO

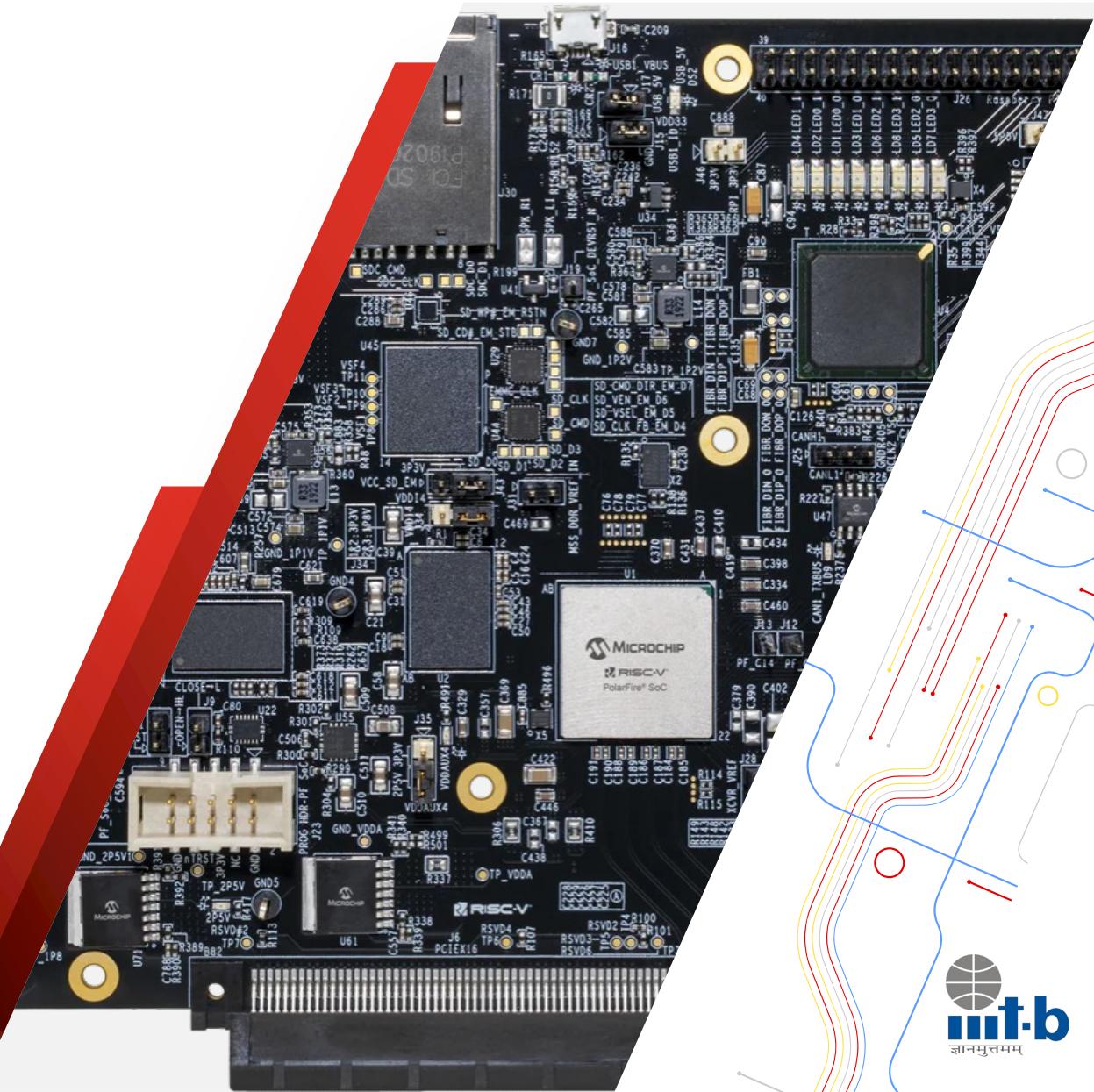
Jan 11, 2023



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Agenda

- Objective
- Algorithms
- CPU implementation
- FPGA Implementation
- Results
- Future work
- Acknowledgements
- References



Objective

- To implement a Hardware-Software co-design to accelerate brain tumour segmentation using a Polar-Fire Soc Icicle kit MPFS250T_ESFCVG484.
- The image segmentation techniques to be implemented are **Water-shed technique** and **Otsu-thresholding technique**.
- **Hardware Software co-design:** The water-shed technique is to be implemented in software (RISC-V CPU) and the otsu-thresholding technique in the hardware (FPGA fabric)

Algorithms



PROCESSING

1. Watershed Segmentation Algorithm

Watershed algorithm is based on extracting sure background and foreground and then using markers will make watershed run and detect the exact boundaries. This algorithm generally helps in detecting touching and overlapping objects in image

2. Otsu Thresholding Algorithm

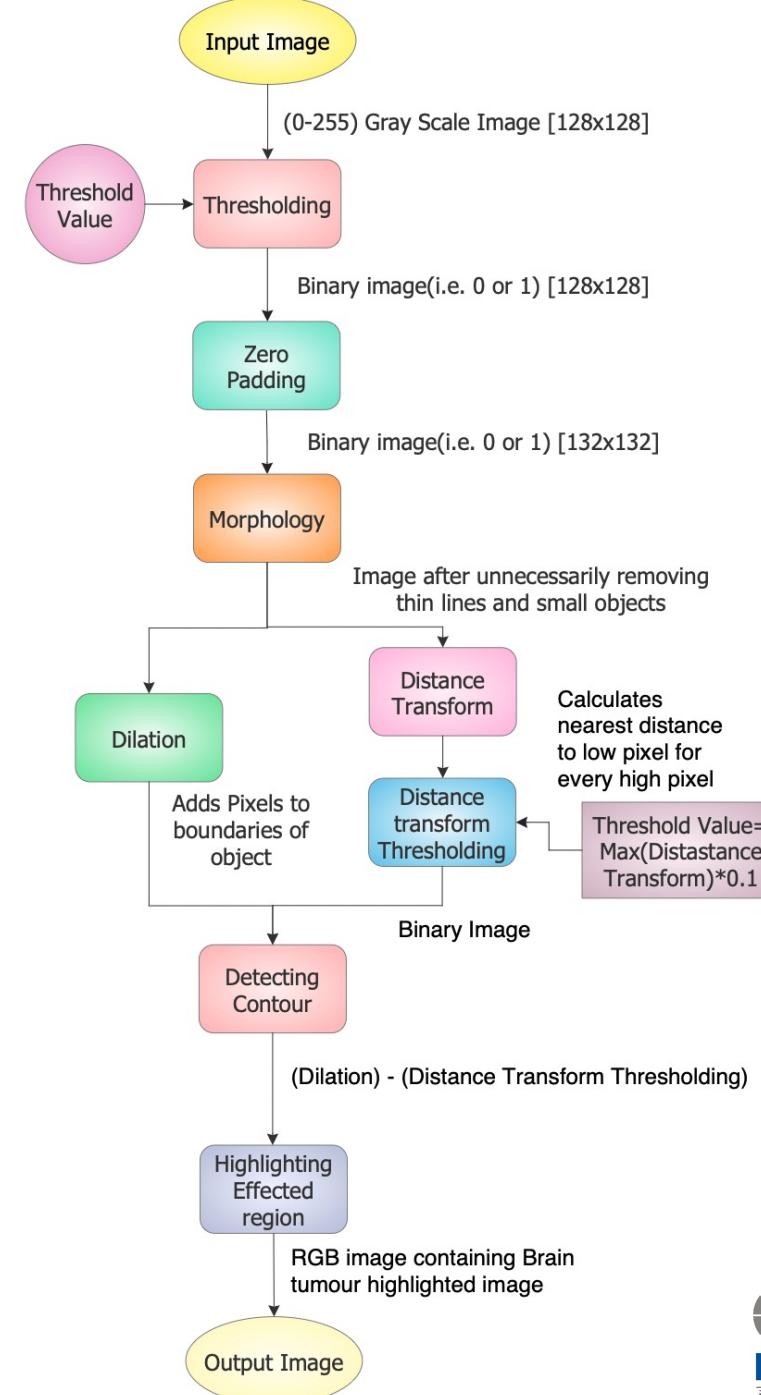
A global adaptive binarization threshold image segmentation algorithm

Software/ CPU implementation

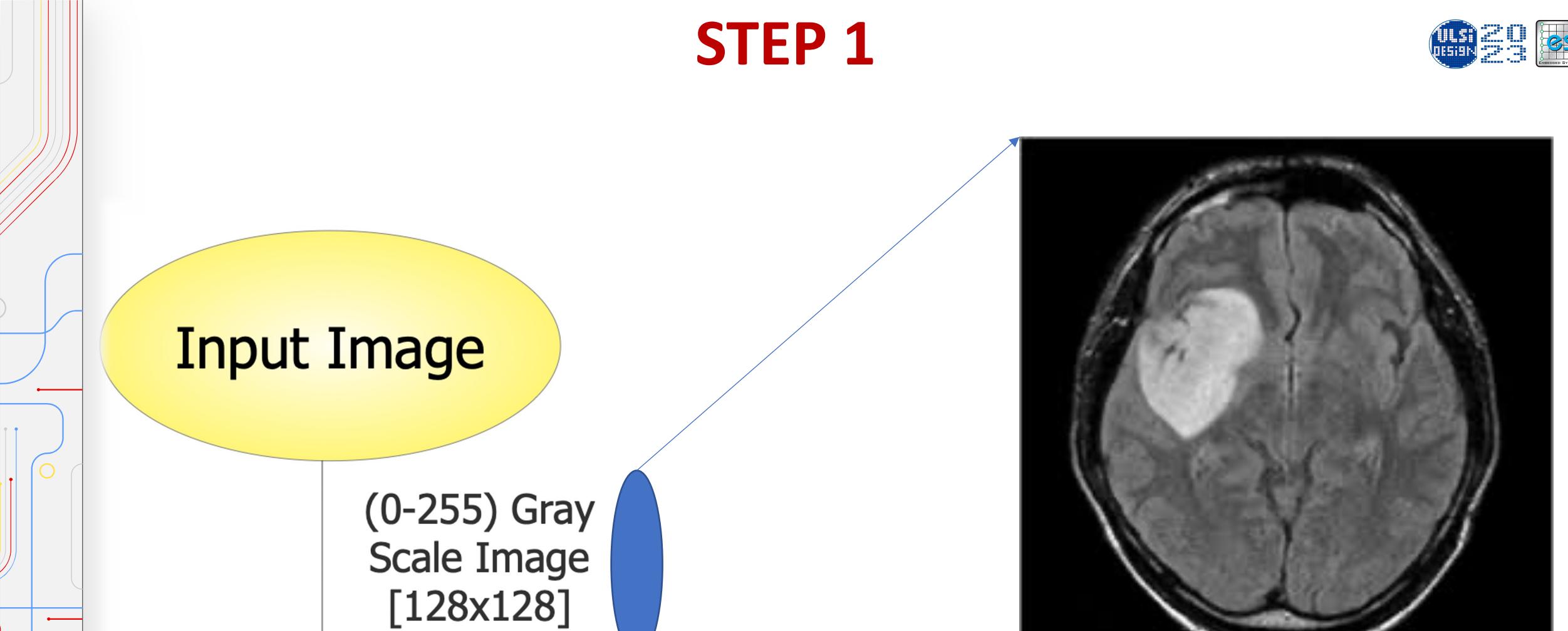
Steps Involved in watershed segmentation:

Watershed segmentation is a region-based segmentation technique that involves the following steps.

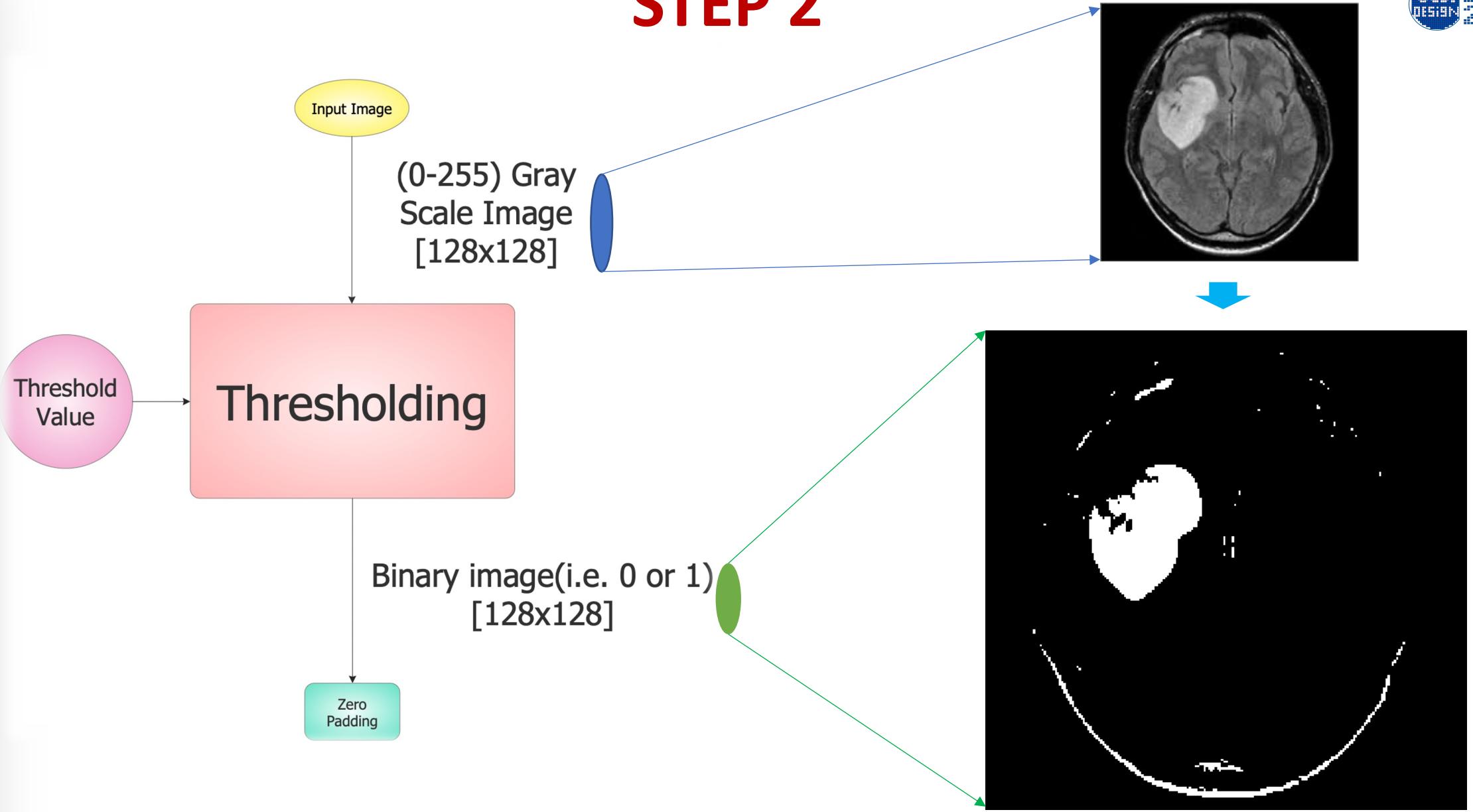
- Threshold Input image
- Morphology
- Dilation
- Distance Transform
- Detecting Contour
- Highlighting Effectuated region in input image



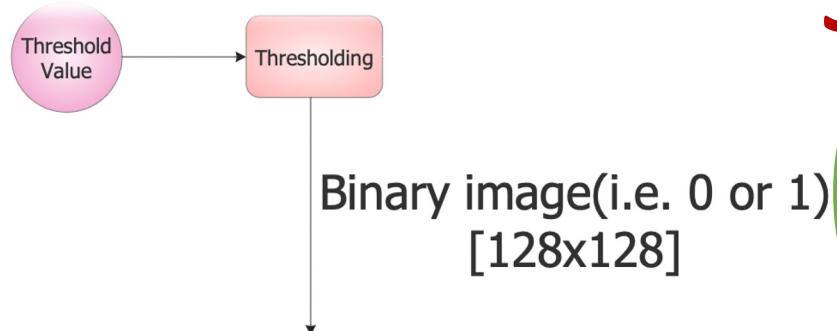
STEP 1



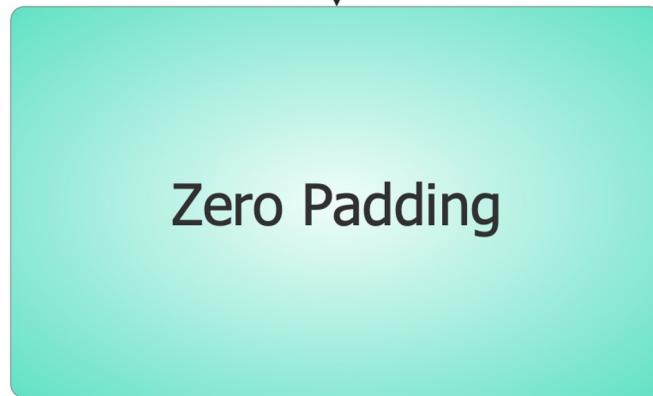
STEP 2



STEP 3

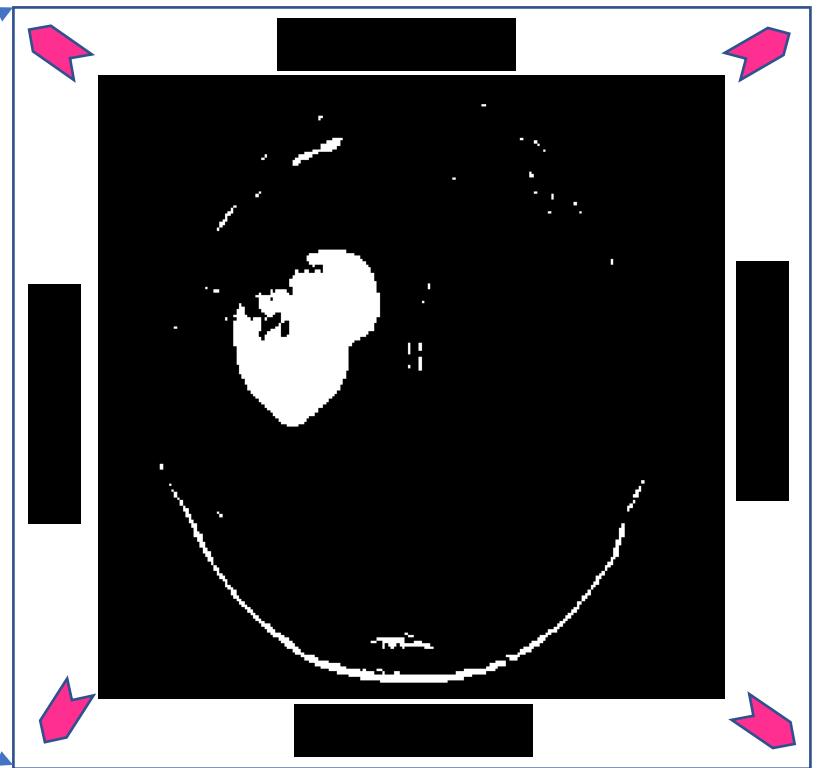


128x128



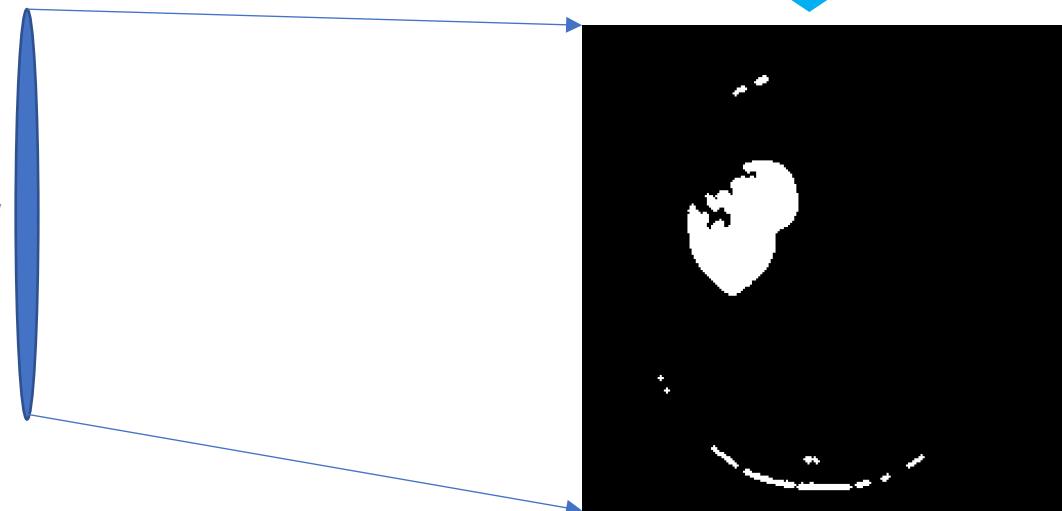
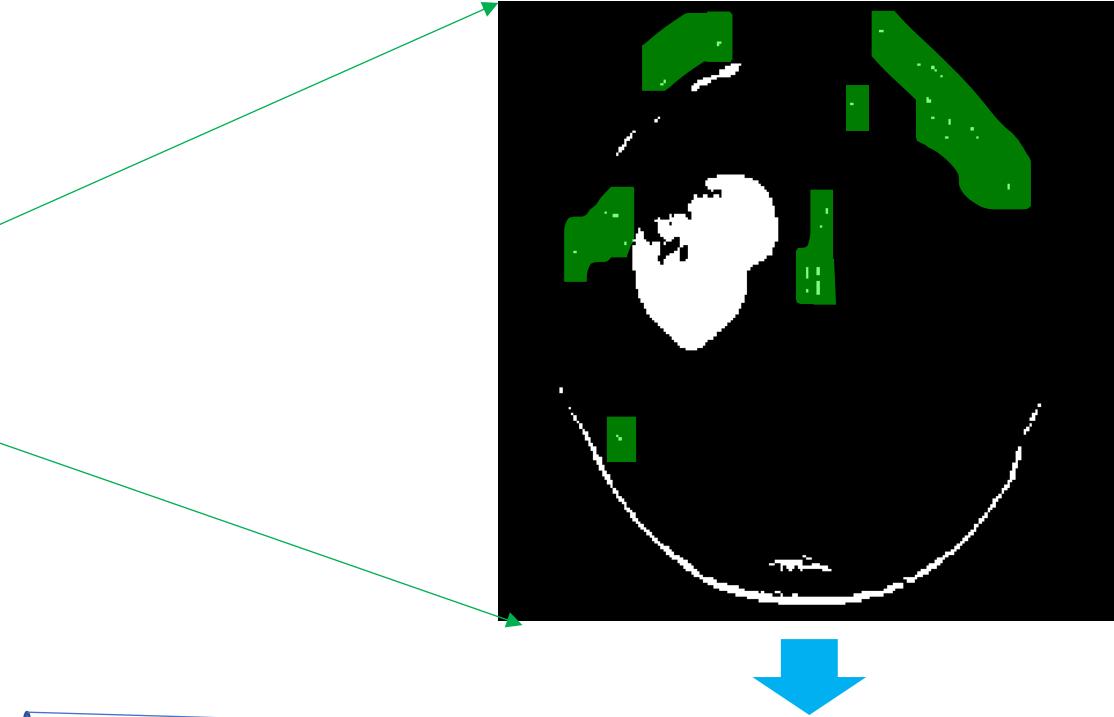
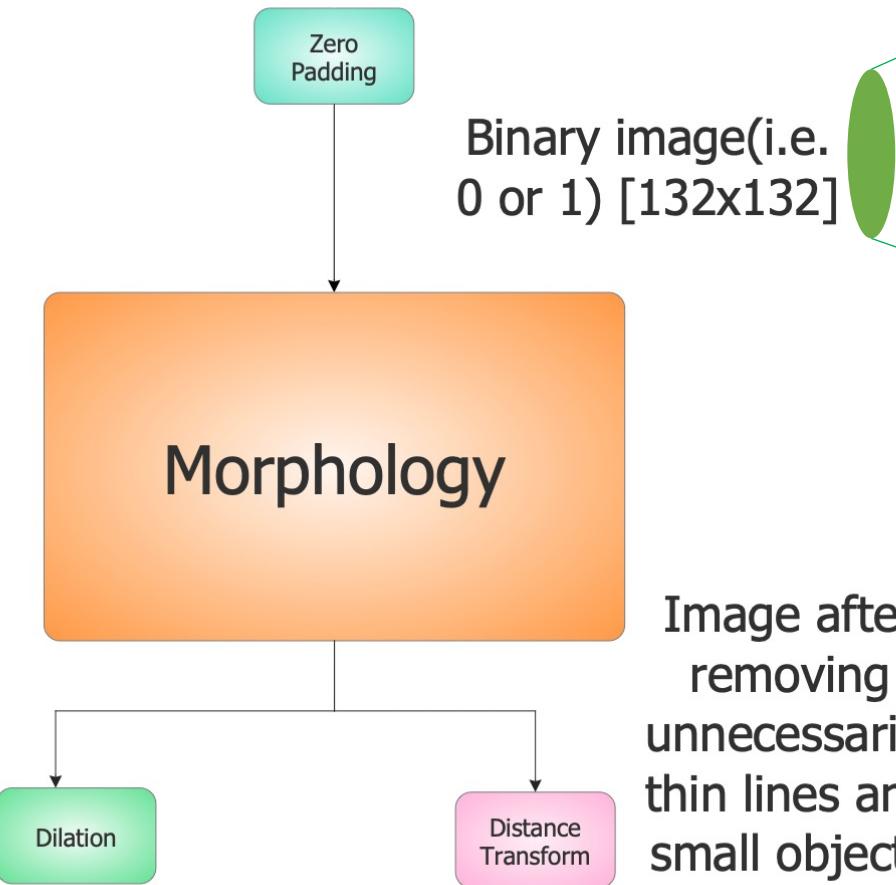
Binary image(i.e. 0 or 1)
[132x132]

Morphology



132x132

STEP 4



STEP 5

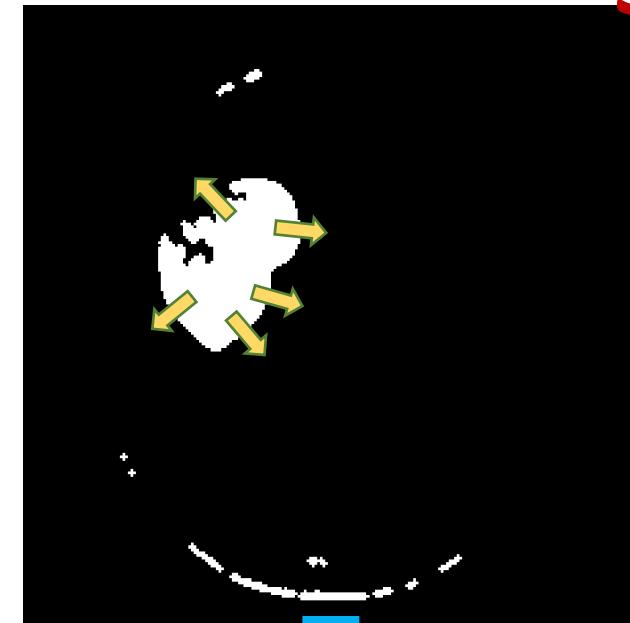


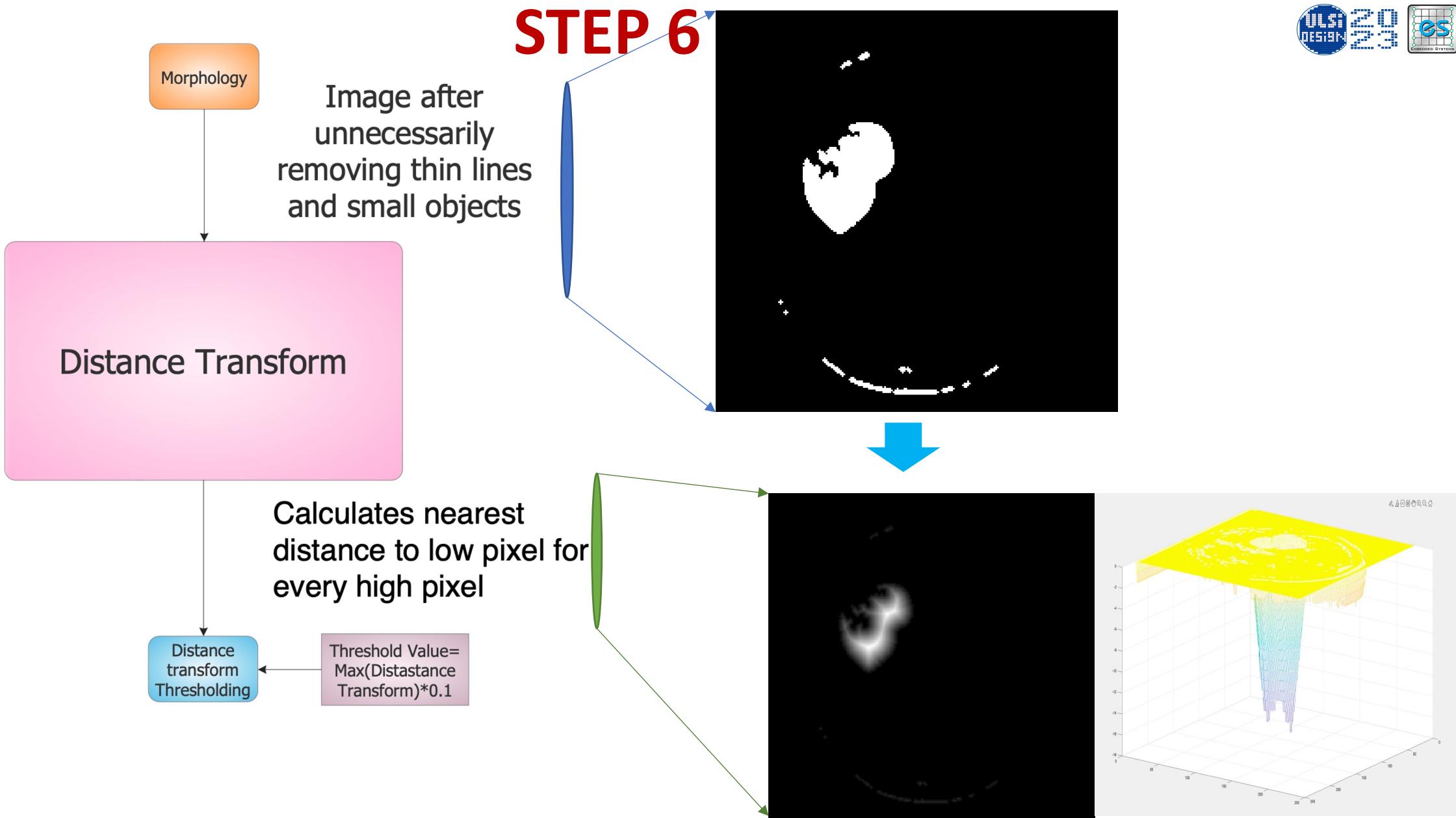
Image after removing unnecessarily thin lines and small objects

Morphology

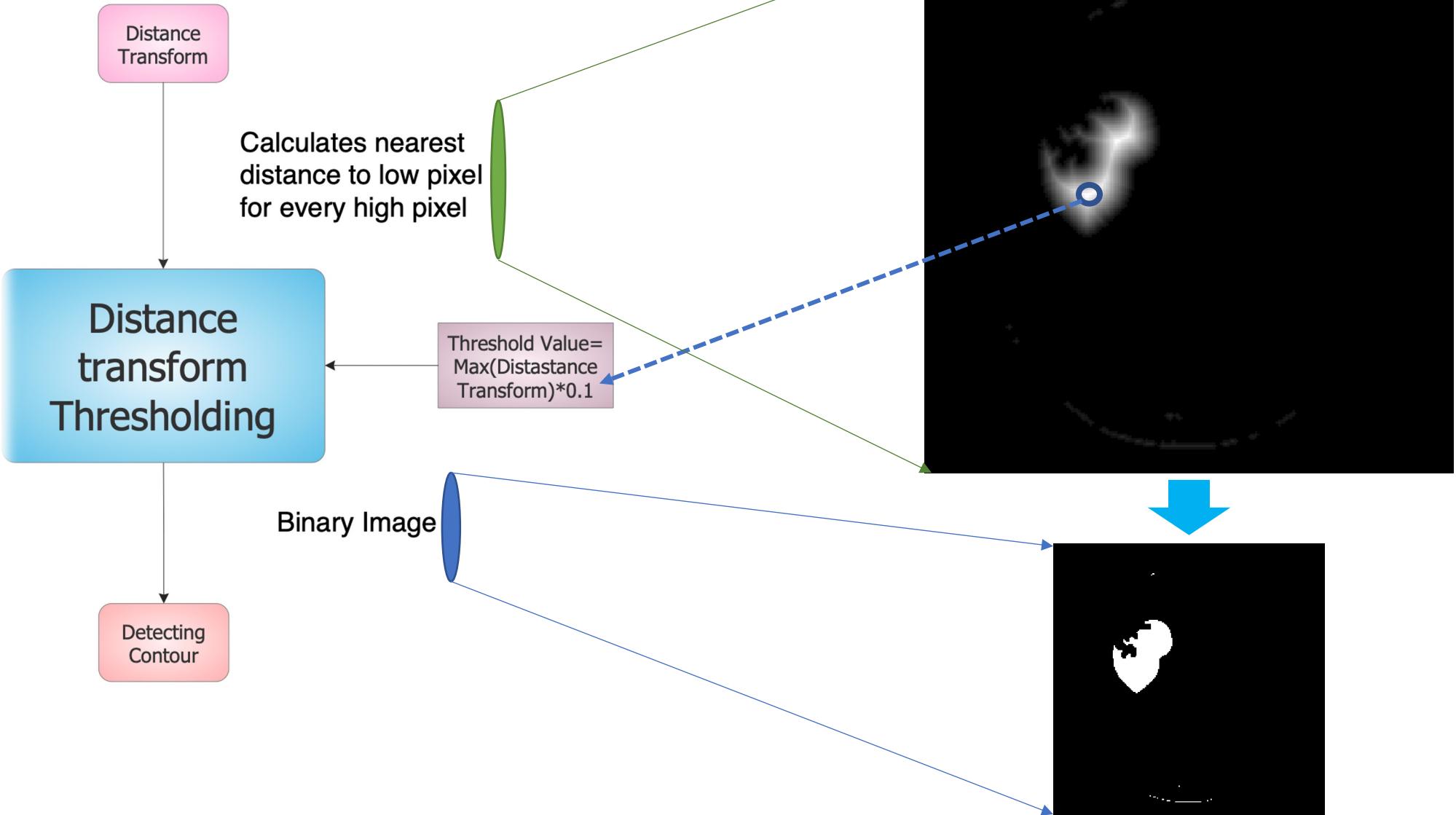
Dilation

Adds Pixels to boundaries of object

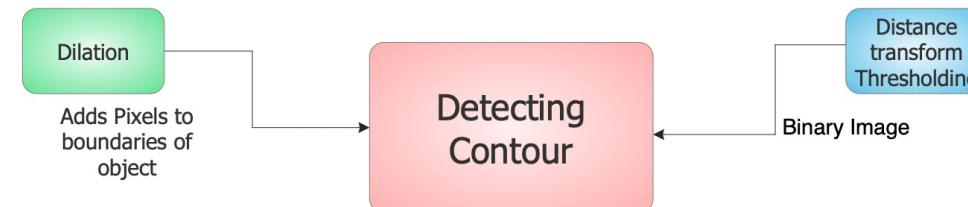
Detecting Contour



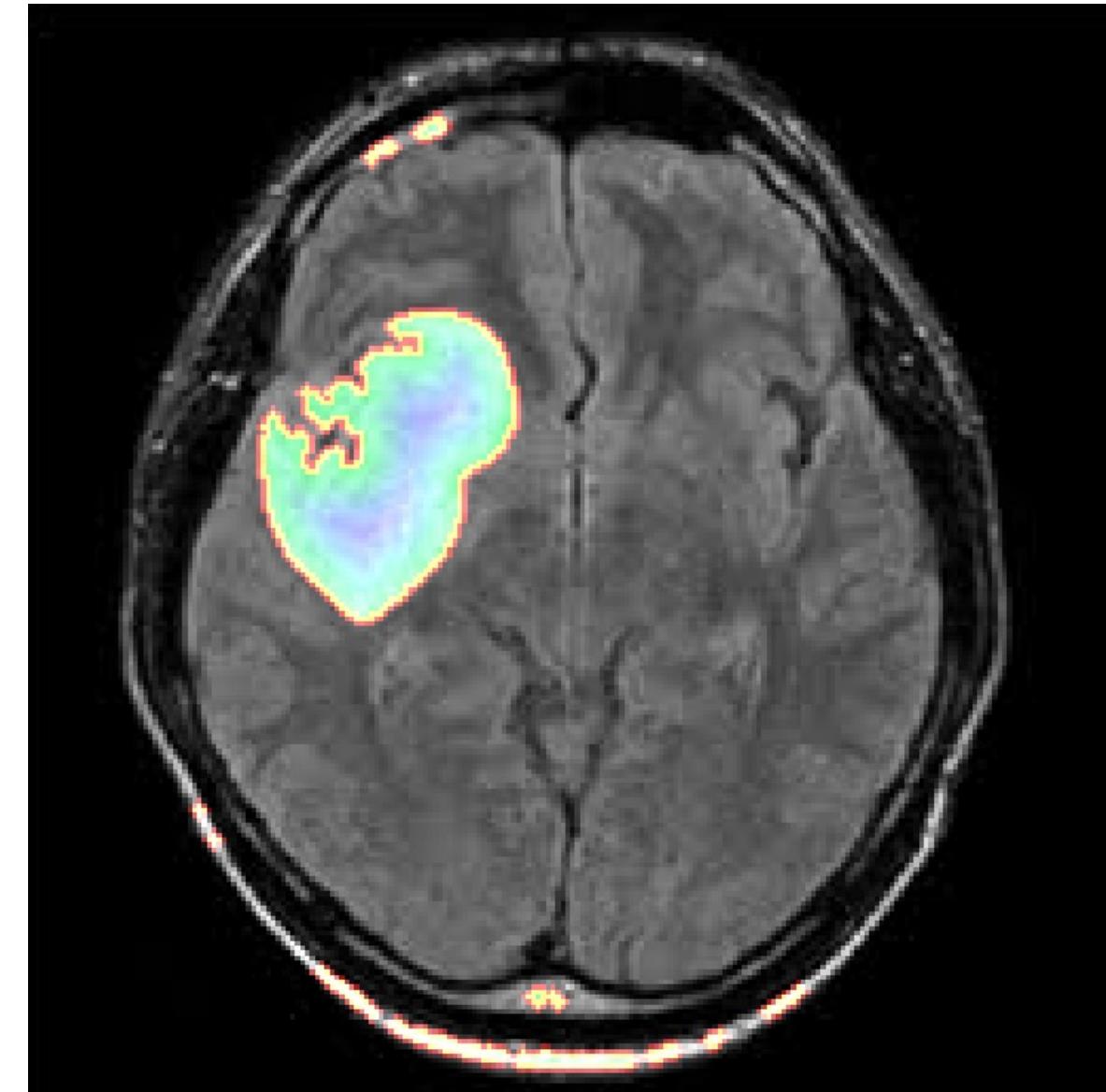
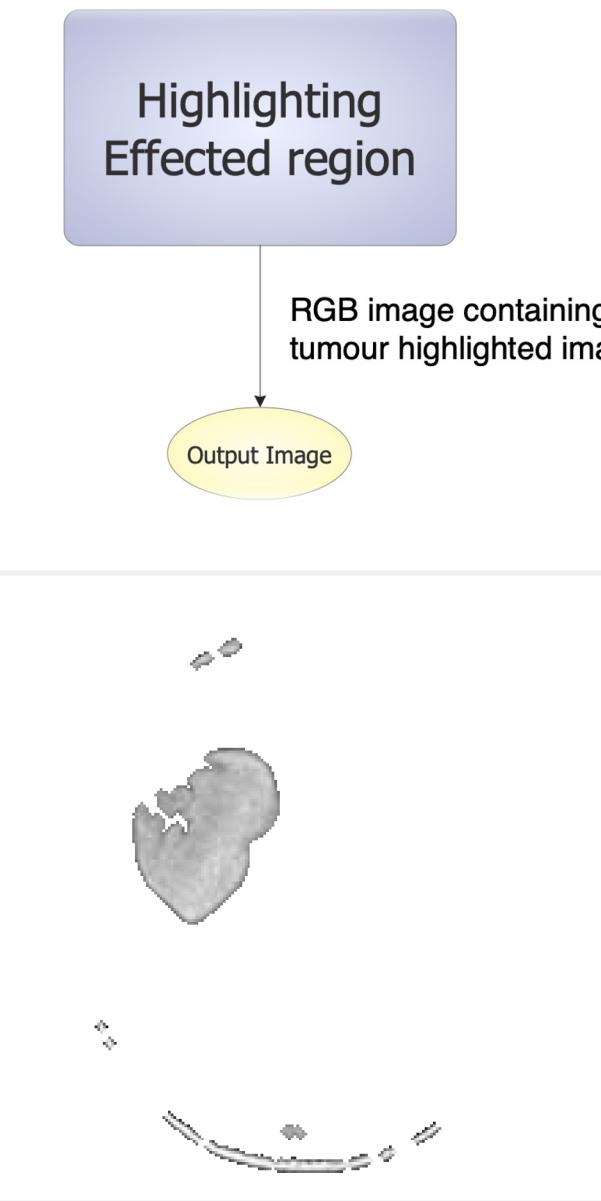
STEP 7



STEP 8



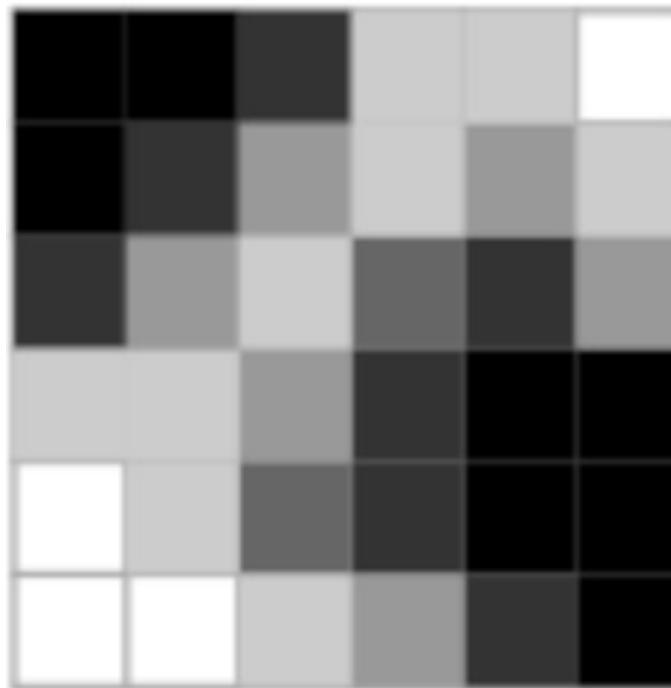
STEP 9



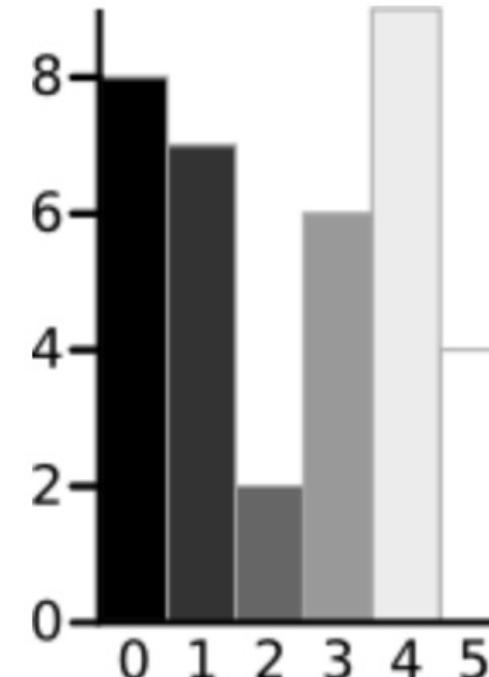
Hardware/FPGA Implementation:

Otsu Thresholding

To perform automatic image thresholding



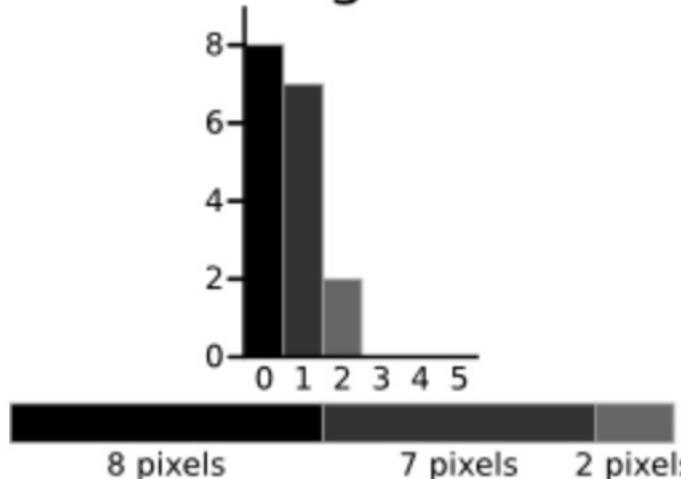
6X6 Image



Histogram

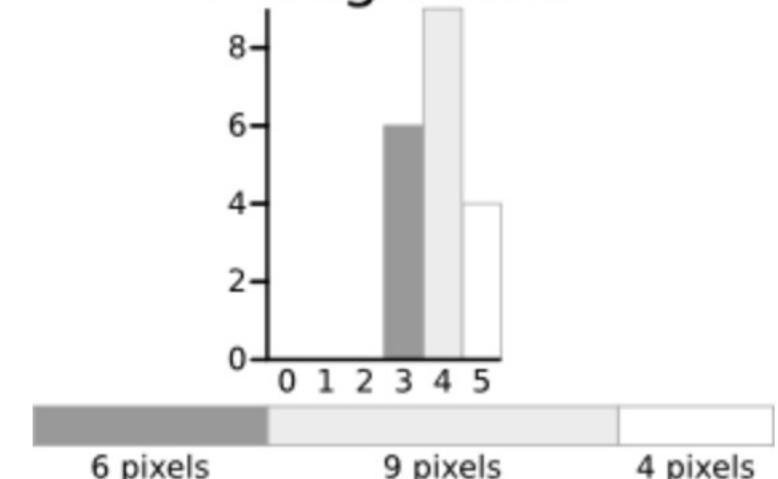
We separate Image as background and Foreground Based on Threshold value

Background



Histogram of Background Image

Foreground

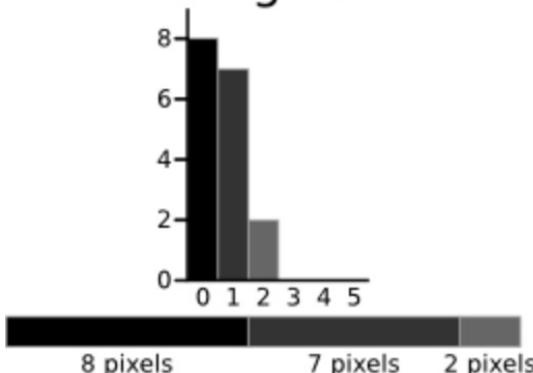


Histogram of Foreground Image

Let Threshold Value=3

Calculation

Background

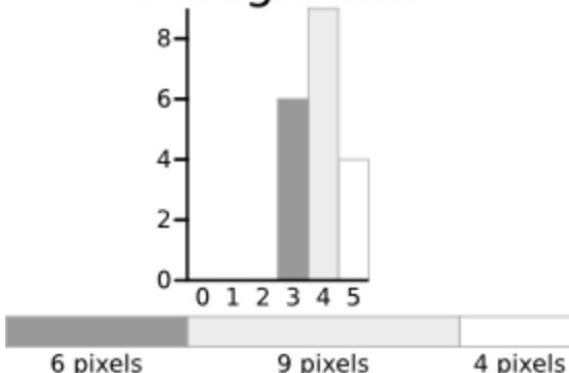


$$\text{Weight } W_b = \frac{8 + 7 + 2}{36} = 0.4722$$

$$\text{Mean } \mu_b = \frac{(0 \times 8) + (1 \times 7) + (2 \times 2)}{17} = 0.6471$$

$$\begin{aligned} \text{Variance } \sigma_b^2 &= \frac{((0 - 0.6471)^2 \times 8) + ((1 - 0.6471)^2 \times 7) + ((2 - 0.6471)^2 \times 2)}{17} \\ &= \frac{(0.4187 \times 8) + (0.1246 \times 7) + (1.8304 \times 2)}{17} \\ &= 0.4637 \end{aligned}$$

Foreground



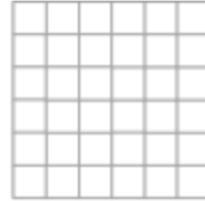
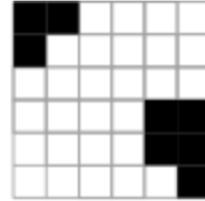
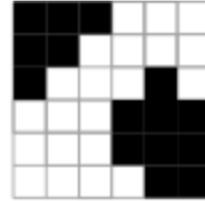
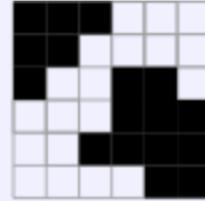
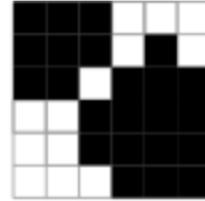
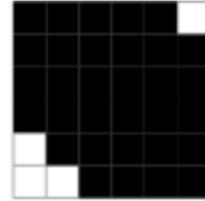
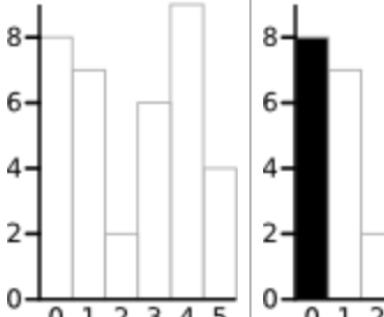
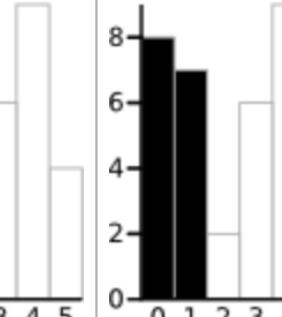
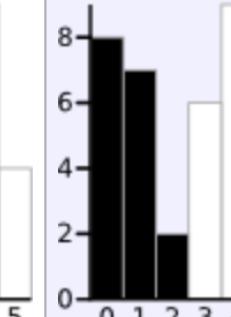
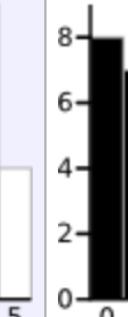
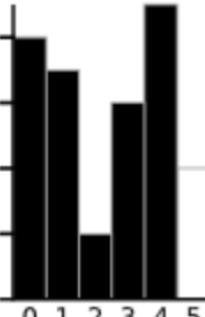
$$\text{Weight } W_f = \frac{6 + 9 + 4}{36} = 0.5278$$

$$\text{Mean } \mu_f = \frac{(3 \times 6) + (4 \times 9) + (5 \times 4)}{19} = 3.8947$$

$$\begin{aligned} \text{Variance } \sigma_f^2 &= \frac{((3 - 3.8947)^2 \times 6) + ((4 - 3.8947)^2 \times 9) + ((5 - 3.8947)^2 \times 4)}{19} \\ &= \frac{(4.8033 \times 6) + (0.0997 \times 9) + (4.8864 \times 4)}{19} \\ &= 0.5152 \end{aligned}$$

$$\begin{aligned} \text{Within Class Variance } \sigma_W^2 &= W_b \sigma_b^2 + W_f \sigma_f^2 = 0.4722 * 0.4637 + 0.5278 * 0.5152 \\ &= 0.4909 \end{aligned}$$

Comparison

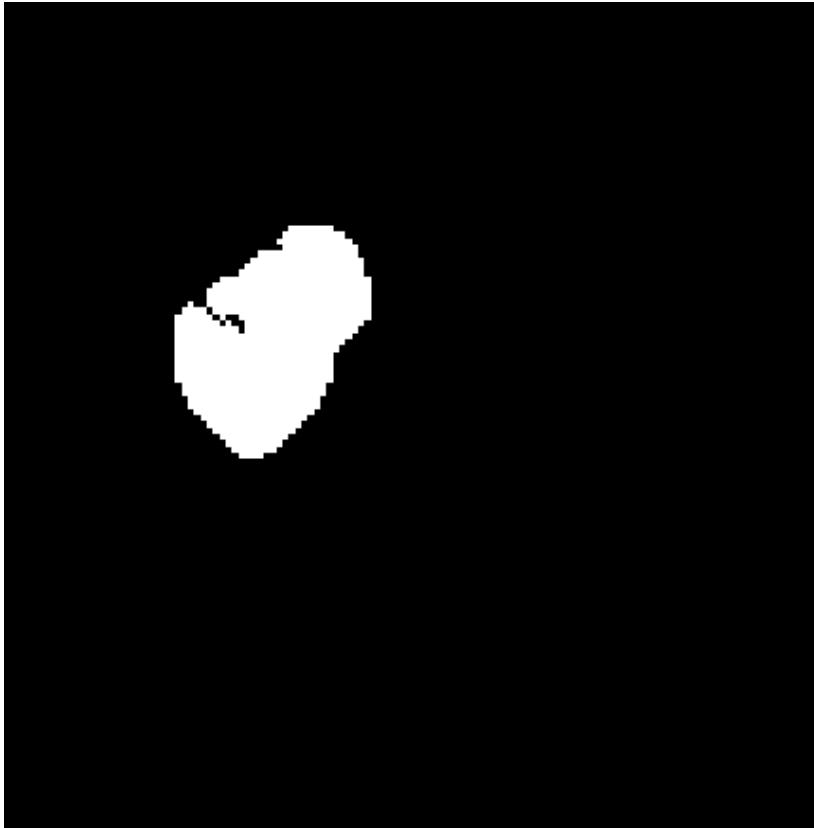
Threshold	T=0	T=1	T=2	T=3	T=4	T=5
						
						
Weight, Background	$W_b = 0$	$W_b = 0.222$	$W_b = 0.4167$	$W_b = 0.4722$	$W_b = 0.6389$	$W_b = 0.8889$
Mean, Background	$M_b = 0$	$M_b = 0$	$M_b = 0.4667$	$M_b = 0.6471$	$M_b = 1.2609$	$M_b = 2.0313$
Variance, Background	$\sigma^2_b = 0$	$\sigma^2_b = 0$	$\sigma^2_b = 0.2489$	$\sigma^2_b = 0.4637$	$\sigma^2_b = 1.4102$	$\sigma^2_b = 2.5303$
Weight, Foreground	$W_f = 1$	$W_f = 0.7778$	$W_f = 0.5833$	$W_f = 0.5278$	$W_f = 0.3611$	$W_f = 0.1111$
Mean, Foreground	$M_f = 2.3611$	$M_f = 3.0357$	$M_f = 3.7143$	$M_f = 3.8947$	$M_f = 4.3077$	$M_f = 5.000$
Variance, Foreground	$\sigma^2_f = 3.1196$	$\sigma^2_f = 1.9639$	$\sigma^2_f = 0.7755$	$\sigma^2_f = 0.5152$	$\sigma^2_f = 0.2130$	$\sigma^2_f = 0$
Within Class Variance	$\sigma^2_w = 3.1196$	$\sigma^2_w = 1.5268$	$\sigma^2_w = 0.5561$	$\sigma^2_w = 0.4909$	$\sigma^2_w = 0.9779$	$\sigma^2_w = 2.2491$

Taken from: <http://www.labbookpages.co.uk/software/imgProc/otsuThreshold.html>

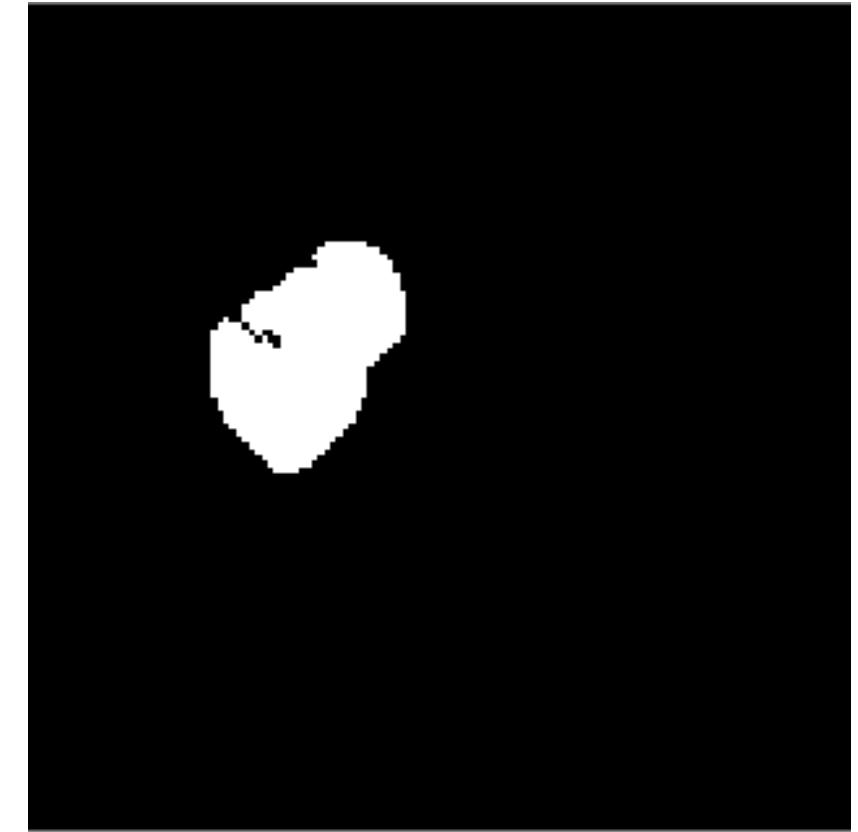
Implementation

- The proposed design is implemented on a Polar-Fire Soc Icicle kit for 128*128 pixels sized brain tumor image.
- The water-shed Algorithm is implemented on one of the five 64-bit RISC-V cores available on the kit. It consumes less than 128kB of the scratch-pad L2 cache memory.
- The Otsu-thresholding algorithm is implemented on the FPGA fabric.
- The UART is used to extract the output data from PS to PC.
- The post-processing of the extracted data is done using MATLAB.
- The Cross-verification of the resulted output images from our implementation is done with the MATLAB generated images for the same input image.

Results of Watershed: Morphology

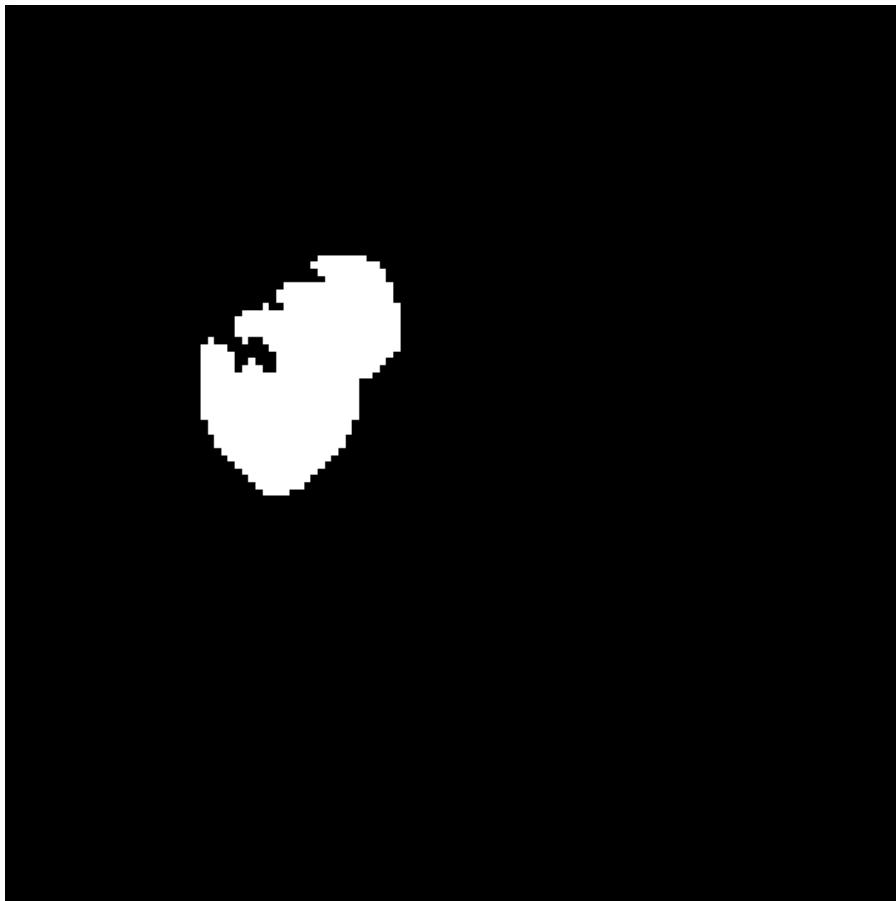


SIMULATED OUTPUT

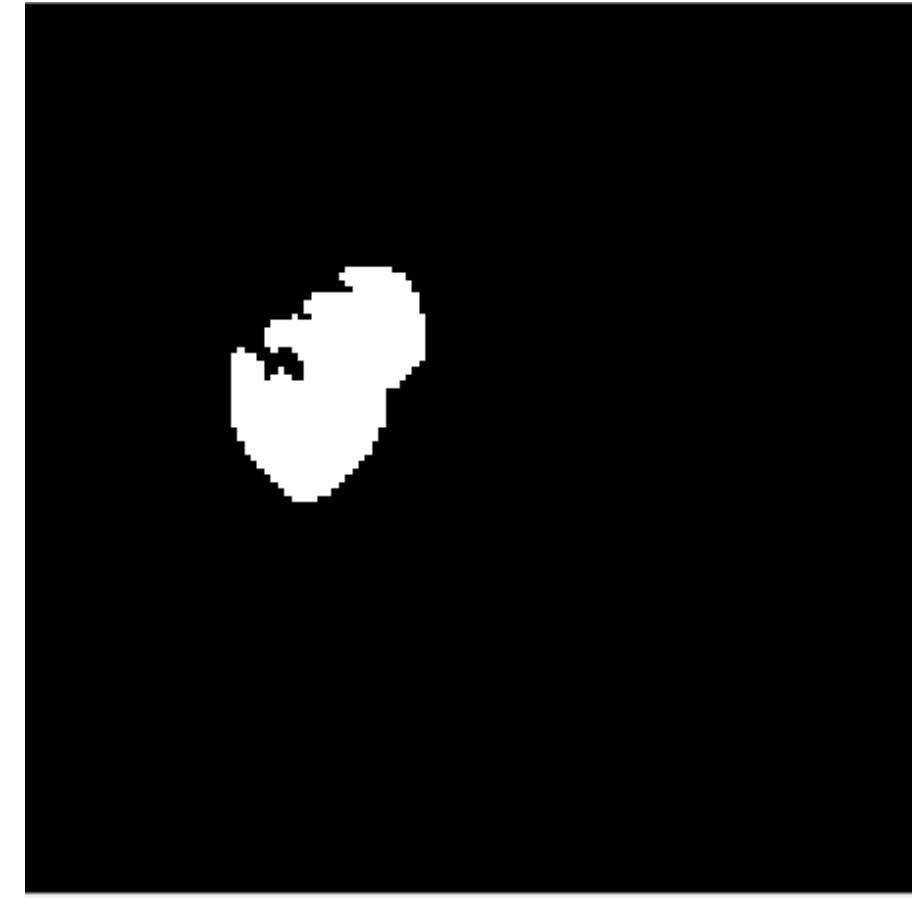


HARDWARE OUTPUT

Results of Watershed: Dilation



SIMULATED OUTPUT

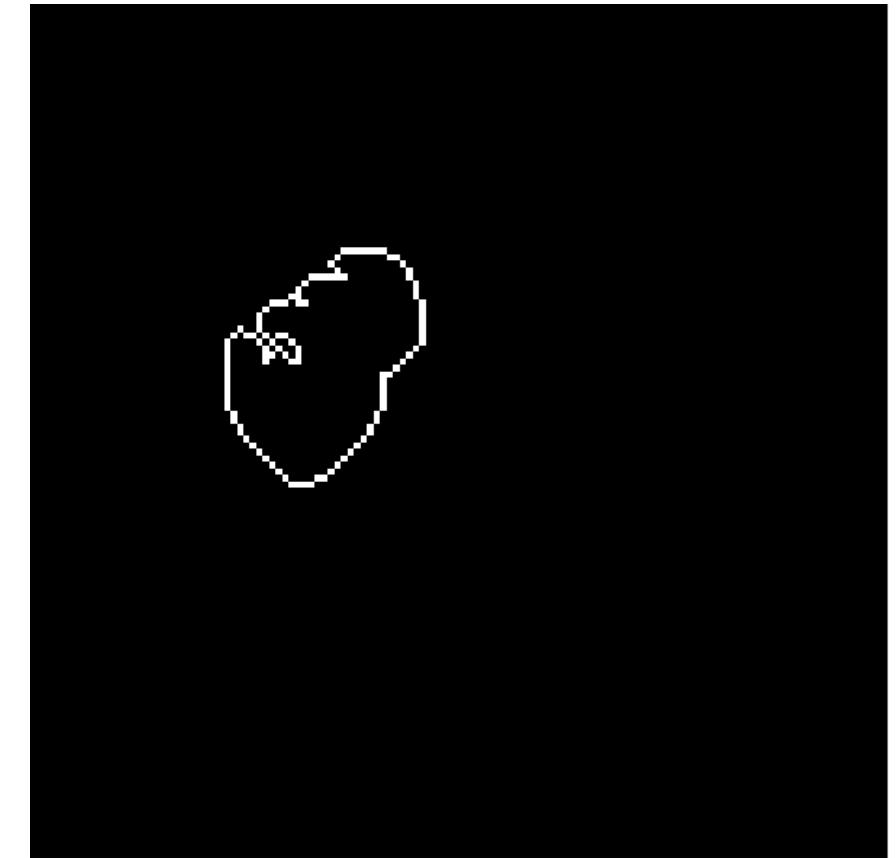


HARDWARE OUTPUT

Results of Watershed: Contour Detection



SIMULATED OUTPUT

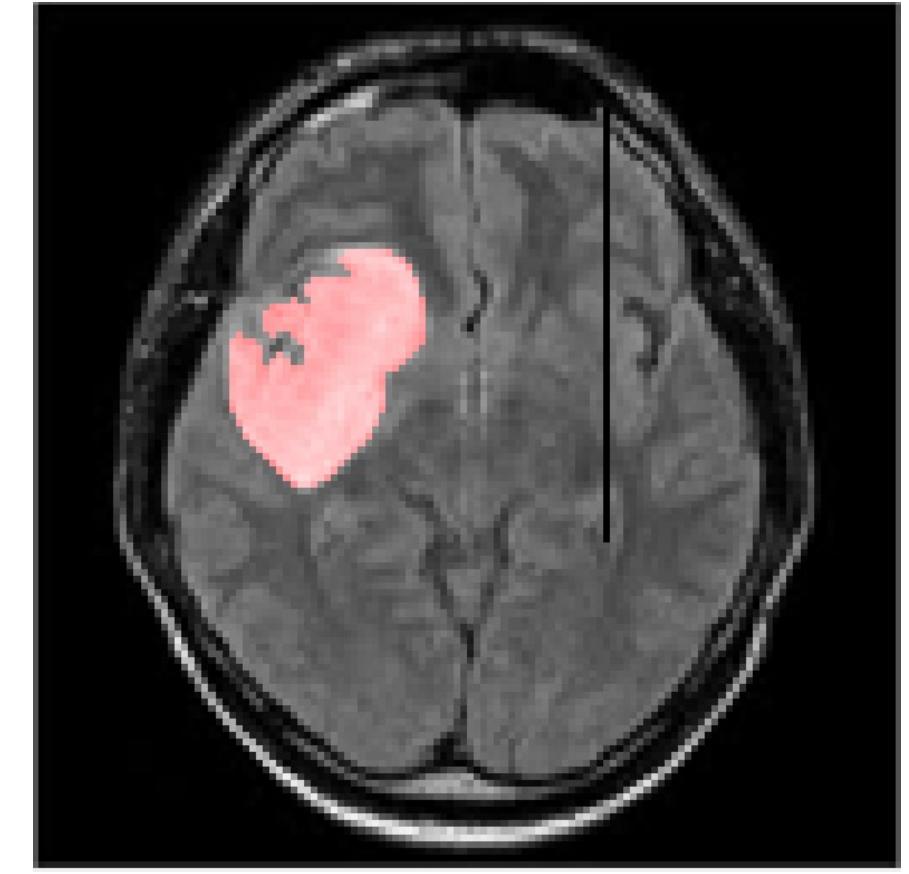


HARDWARE OUTPUT

Results of Watershed: Highlighted region

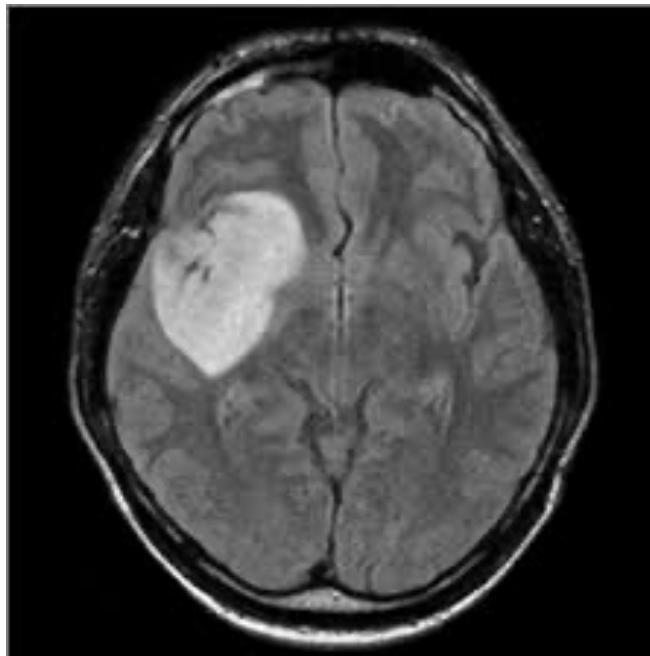


SIMULATED OUTPUT

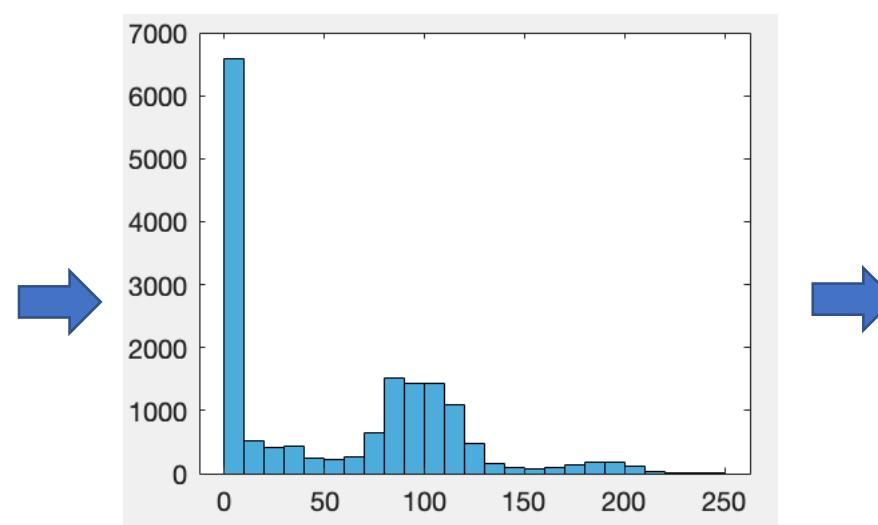


HARDWARE OUTPUT

Results of Otsu Thresholding:



Input Image



Histogram of Input Image



Thresholded
Image using
Otsu's Method

Results: Logic Resources

Detailed Logic Resource Usage

Type	4LUT	DFF
Fabric Logic	18549	21201
uSRAM Interface Logic	0	0
LSRAM Interface Logic	216	216
Math Interface Logic	756	756
Total Used	19521	22173

Resource Usage

Type	Used	Total	Percentage
4LUT	19521	254196	7.68
DFF	22173	254196	8.72
I/O Register	0	432	0.00
User I/O	101	144	70.14
-- Single-ended I/O	101	144	70.14
-- Differential I/O Pairs	0	72	0.00
uSRAM	0	2352	0.00
LSRAM	6	812	0.74
Math	21	784	2.68
H-Chip Global	1	48	2.08
PLL	0	8	0.00
DLL	0	8	0.00
Transceiver Lanes	0	4	0.00
Transceiver PCIe	0	2	0.00
MSS	0	1	0.00

Results: Power and Timing

Power Summary

	Power (mW)	Percentage
Total Power	166.330	100.0%
Static Power	155.788	93.7%
Dynamic Power	10.542	6.3%

===== 2. Timing Result of HLS-generated IP Core (top-level module: ostu_top) =====

Clock Domain	Target Period	Target Fmax	Worst Slack	Period	Fmax
clk	10.000 ns	100.000 MHz	3.213 ns	6.787 ns	147.341 MHz

Future Work

Classification based on CNN model can be done to auto-detect and classify the tumour. This can be implemented as a hardware-software co-design with tasks (layers) divided between the CPU and FPGA fabric

Acknowledgements:

- Microchip technical support team for multiple Query meetings



References:

- Microchip Smart HLS Guide
- Microchip Libero Guide
- Microchip Soft console Guide
- Polar fire Icicle kit examples:
<https://github.com/polarfire-soc/icicle-kit-reference-design>

THANK YOU!

