

# Object Boundary Refinement Using Level Sets

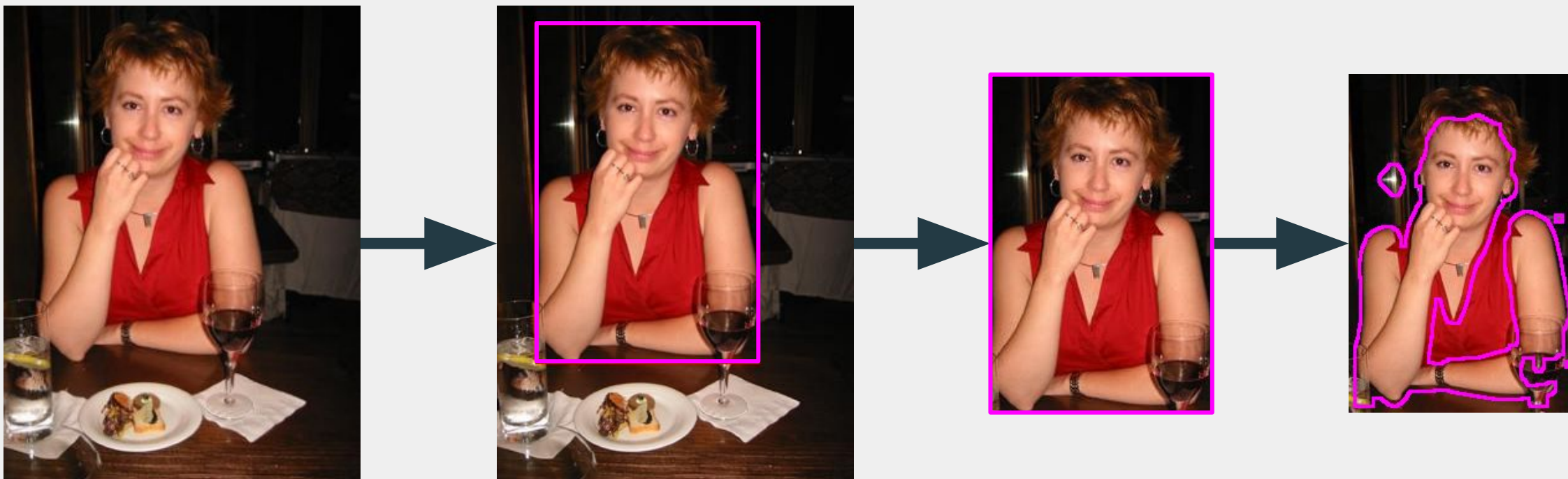
CS 554 - Computer Vision Project Presentation

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# Introduction



# Active Contour Models (Snakes)

Minimizes energy functional:

$$E_{snake} = \int_{\Gamma} (\alpha |\mathbf{v}'(s)|^2 + \beta |\mathbf{v}''(s)|^2 - I(\mathbf{v}(s))) ds$$

Internal Energy

External Energy

**Challenges:**

- Noise
- Initialization.

# Level Set Method

Contour evolution governed by:

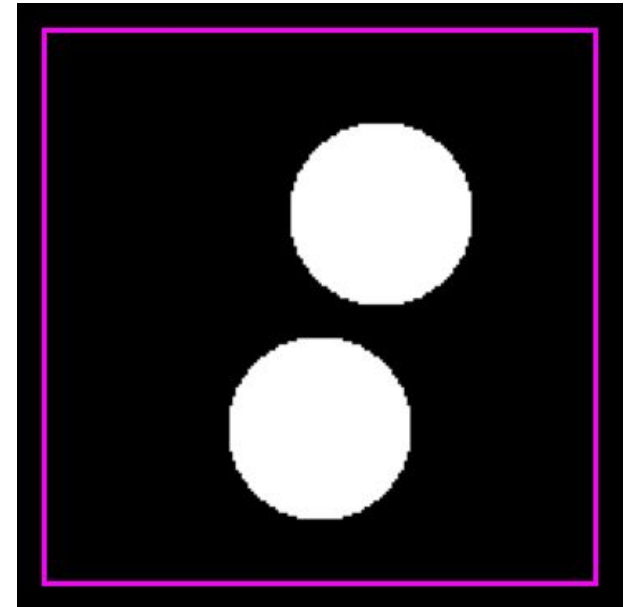
$$\frac{\partial \phi}{\partial t} = \delta(\phi(x)) \cdot (\alpha \nabla \phi \cdot \nabla I + \beta I)$$

**Advantages:**

- Topological changes

**Challenges:**

- Computation
- Noise



# Enhanced Level Set with Sobel Filters

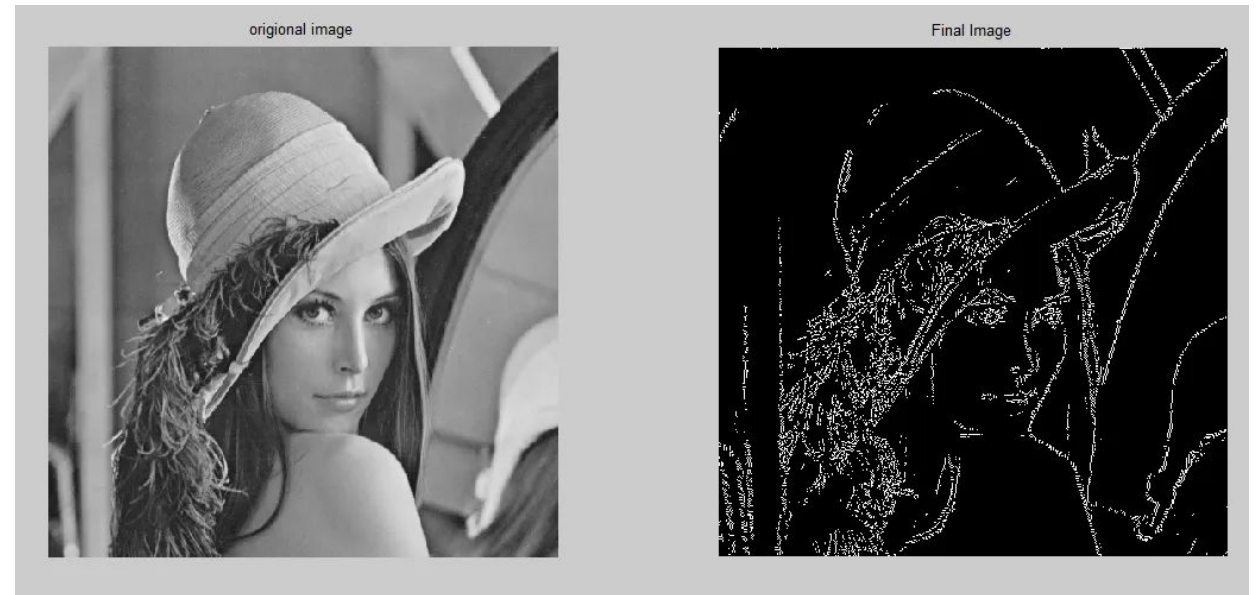
Sobel filter:

-1	0	+1
-2	0	+2
-1	0	+1

Gx

+1	+2	+1
0	0	0
-1	-2	-1

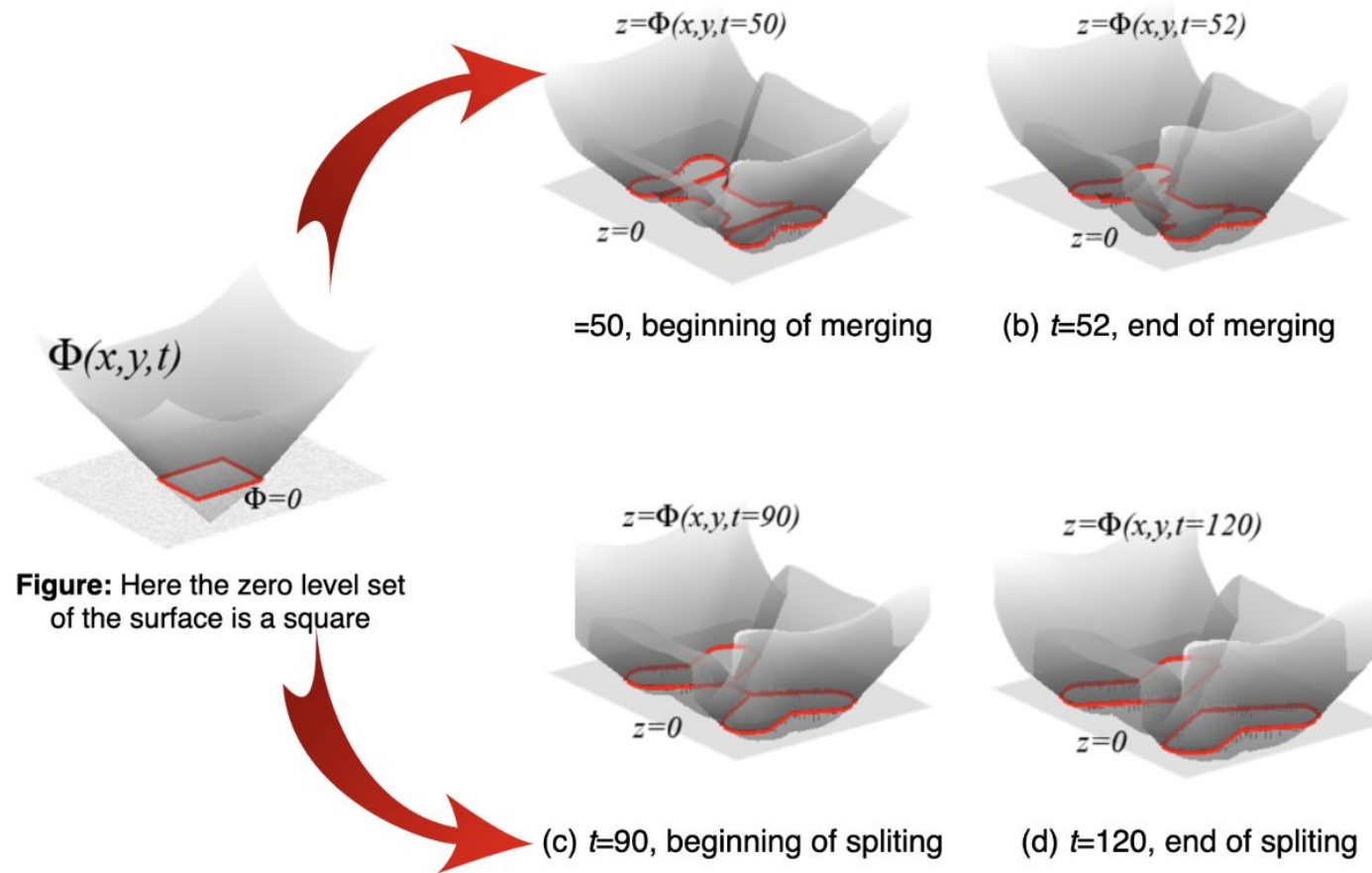
Gy



**Advantages:**

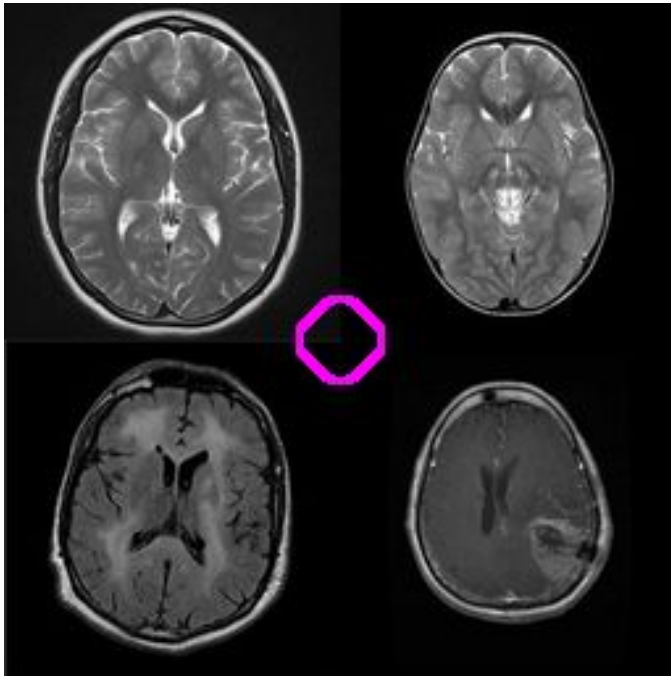
- Noise
- Edge

# Contour Initialization Strategies

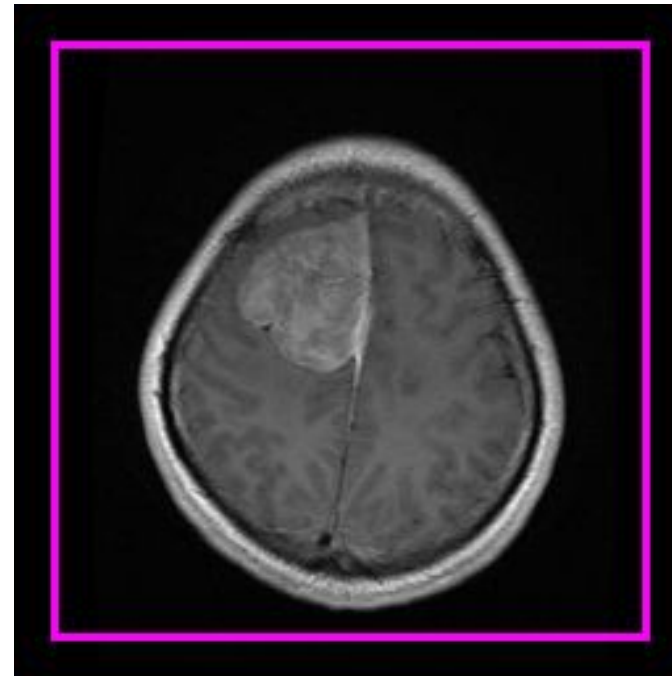


# Contour Initialization Strategies

(a) Default initial boundary  
initialization - mode 1



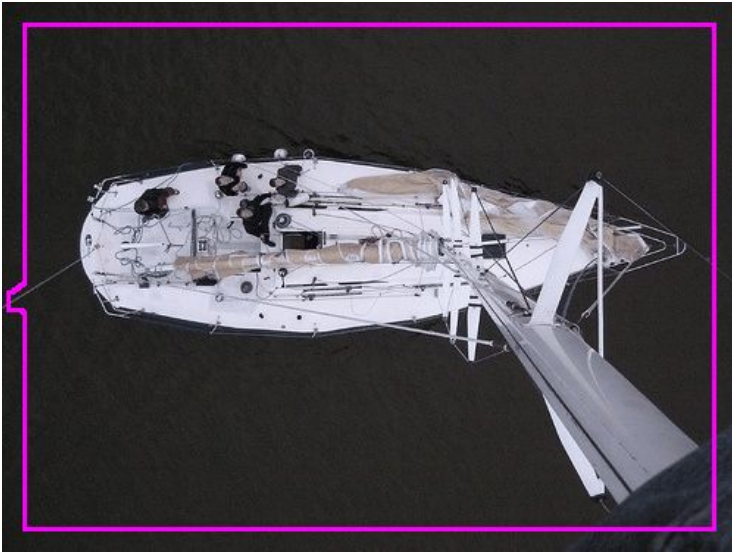
(b) Default initial boundary  
initialization - mode 2



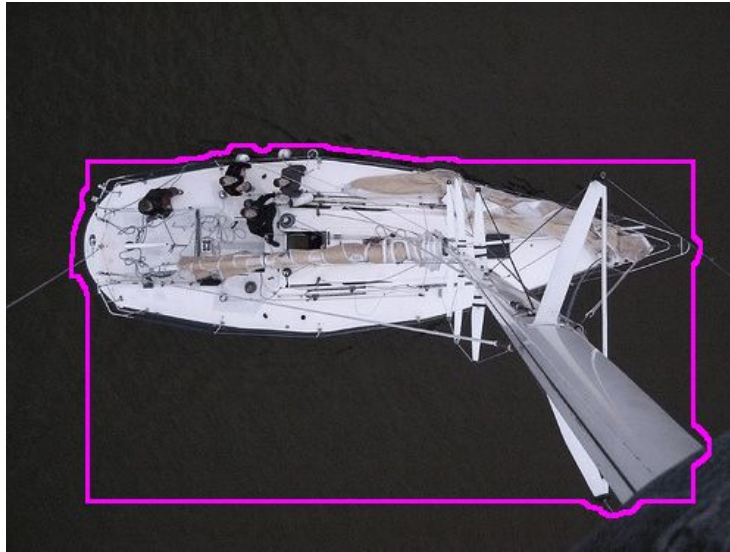


# Contour Initialization Strategies

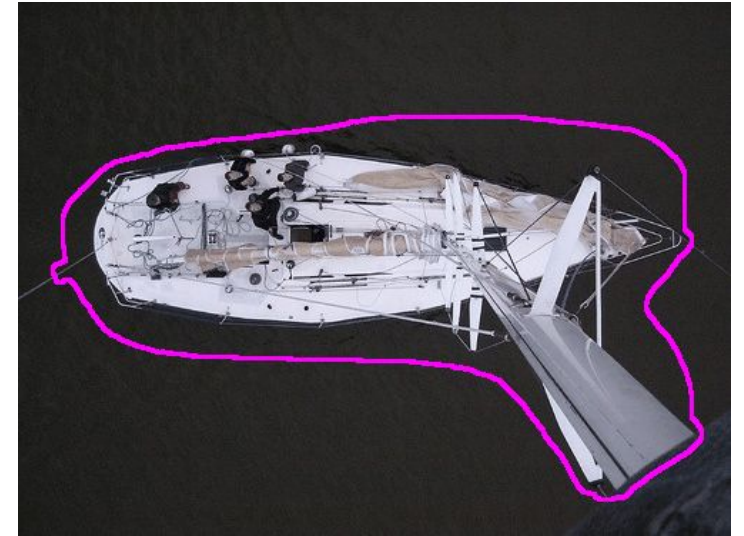
(a) Default initial boundary initialization



(b) user-defined rectangle boundary initialization



(c) user-defined free-form boundary initialization





# Contour Initialization Strategies

## *XGBoost Model Training and Feature Extraction*

**Dataset Used:** Pascal VOC

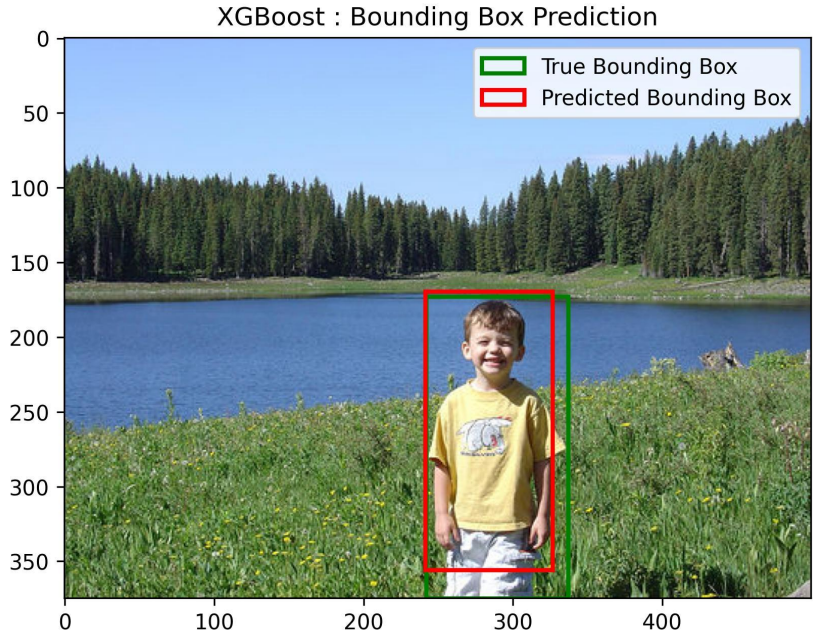
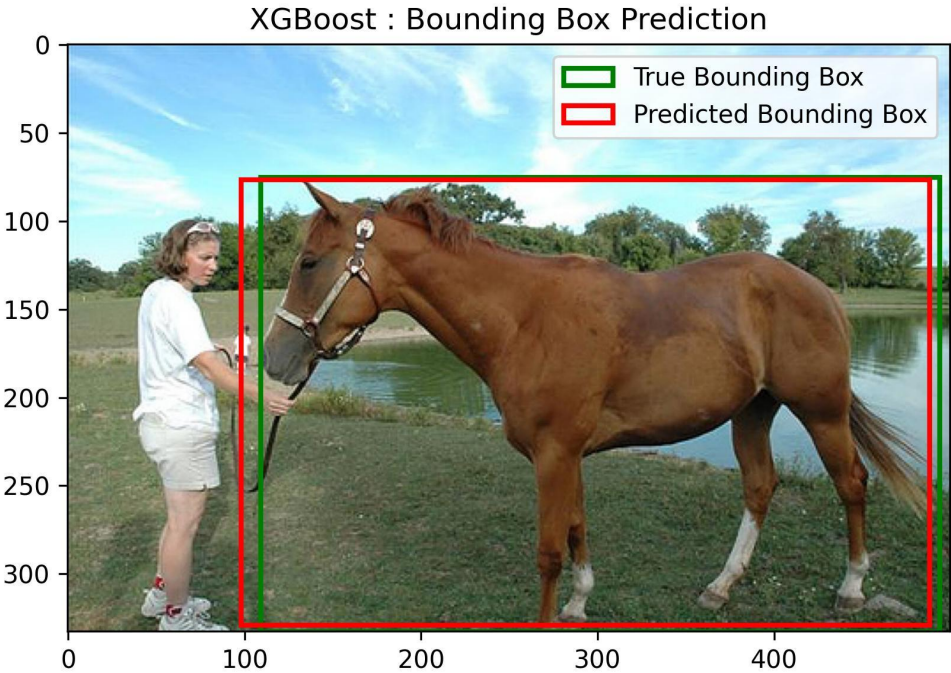
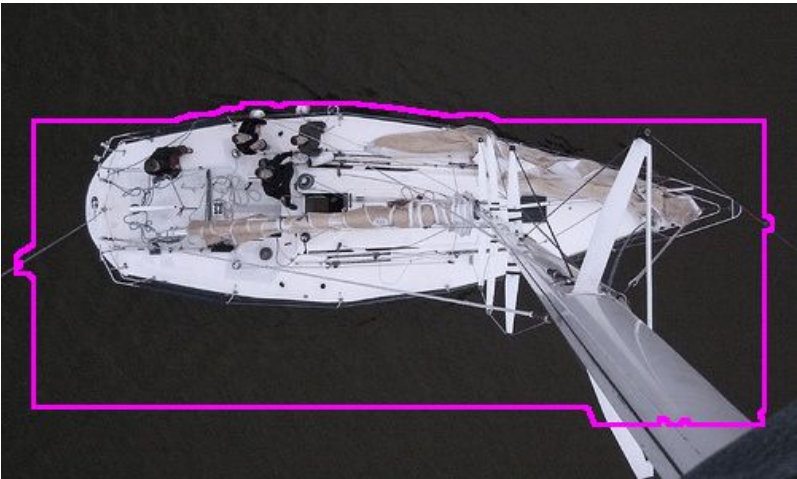
**Purpose:** Train the model to predict bounding boxes for level set initialization.

**Feature Extraction:** Pixel Intensity + HOG Features

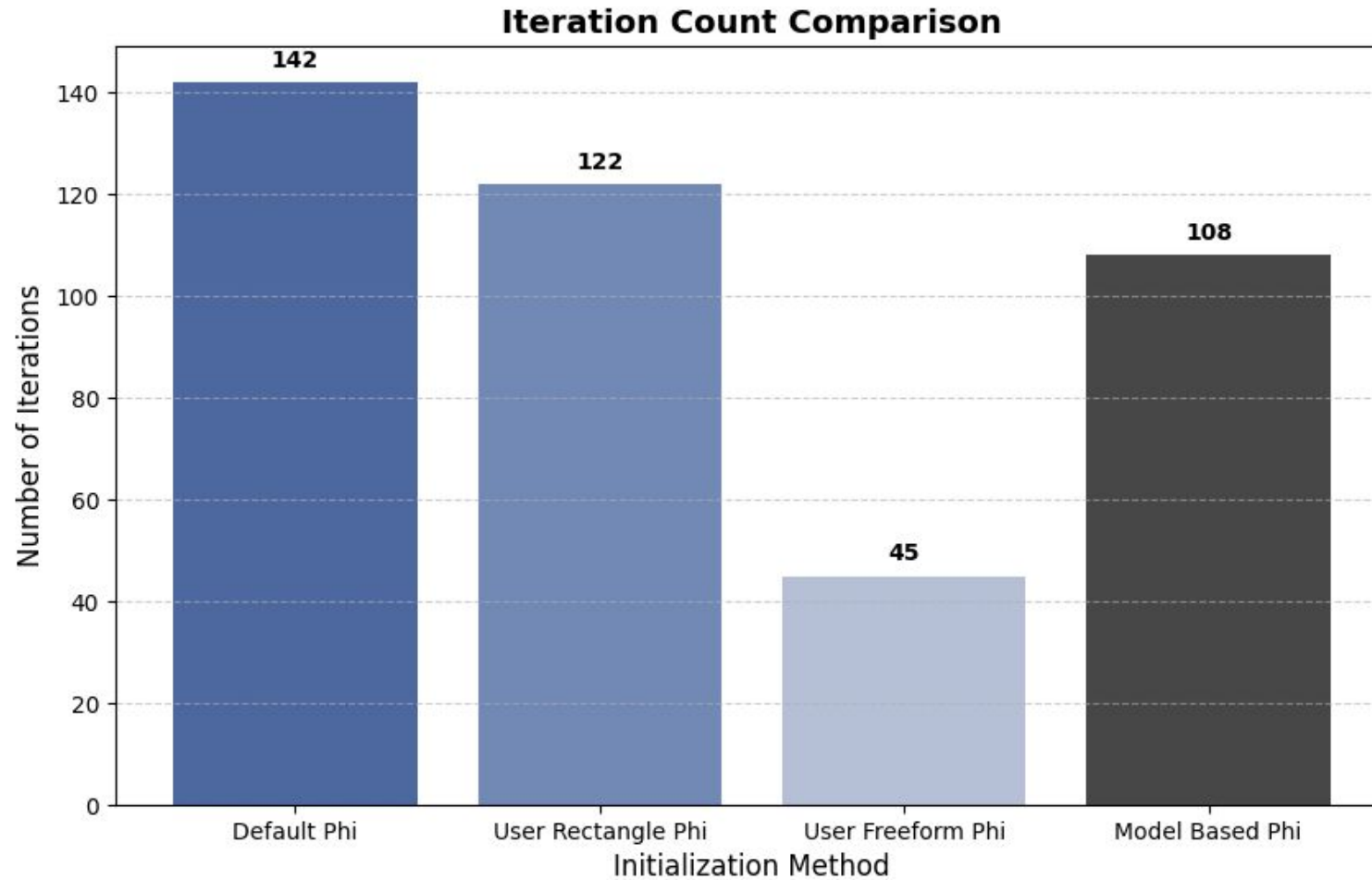
**Target:** Bounding box coordinates [xmin,ymin,xmax,ymax]

# Contour Initialization Strategies

*(d) Predicting initial bounding box using XGBoost*



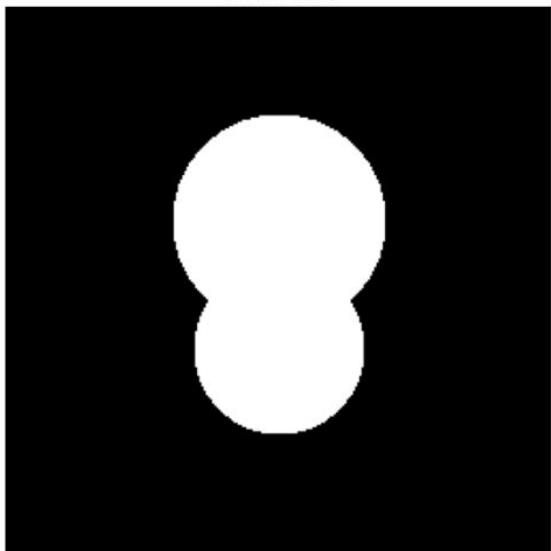
# Contour Initialization Strategies



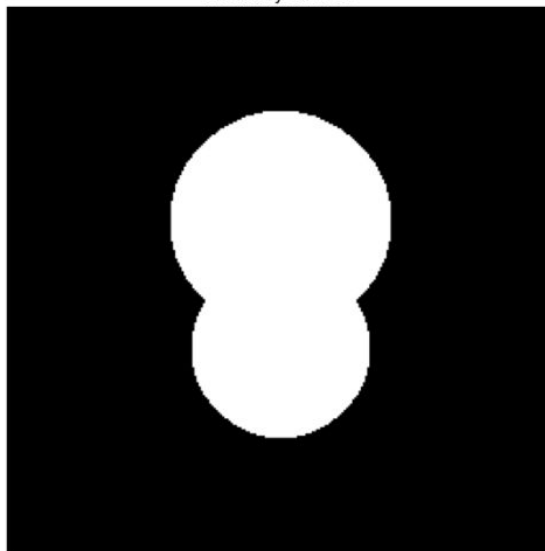
# Results

Segmentation Result after applying the old level set, the Sobel level set and snakes.

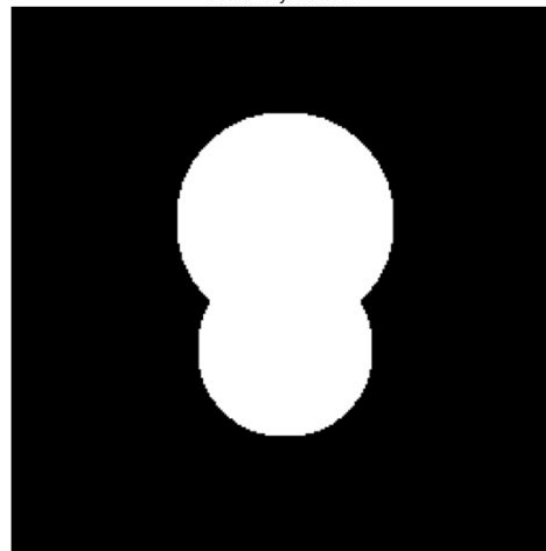
Ground Truth



Simple Level Set Segmentation  
IoU: 0.9406  
Accuracy: 0.9890



New Level Set Segmentation  
IoU: 0.9563  
Accuracy: 0.9920



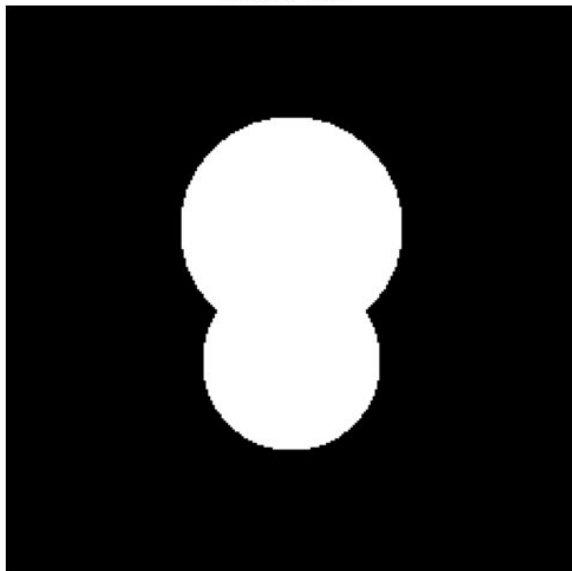
Snake Model Segmentation  
IoU: 0.7289  
Accuracy: 0.9358



# Results

## Adding different level of noise and segmenting using the old level set

Ground Truth



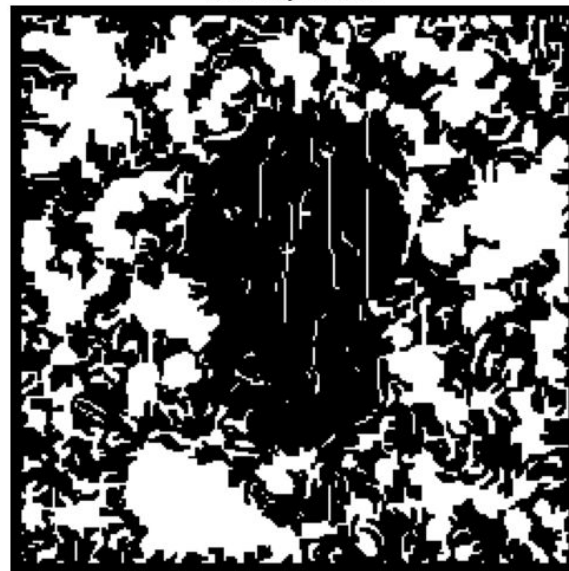
Light Noise Segmentation

IoU: 0.0077  
Accuracy: 0.7853



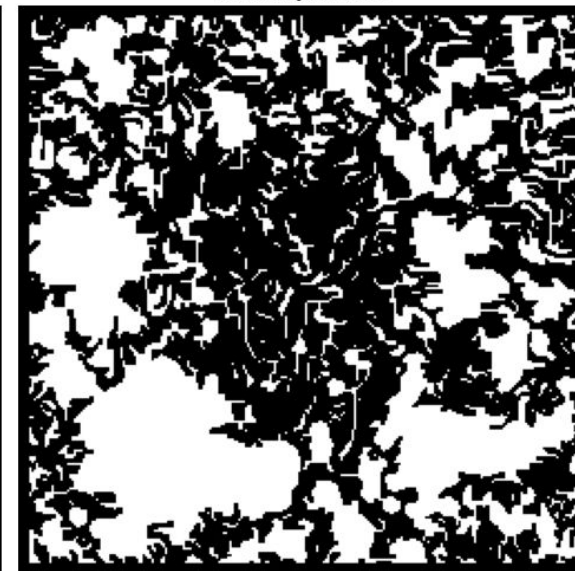
Medium Noise Segmentation

IoU: 0.0158  
Accuracy: 0.4720



Heavy Noise Segmentation

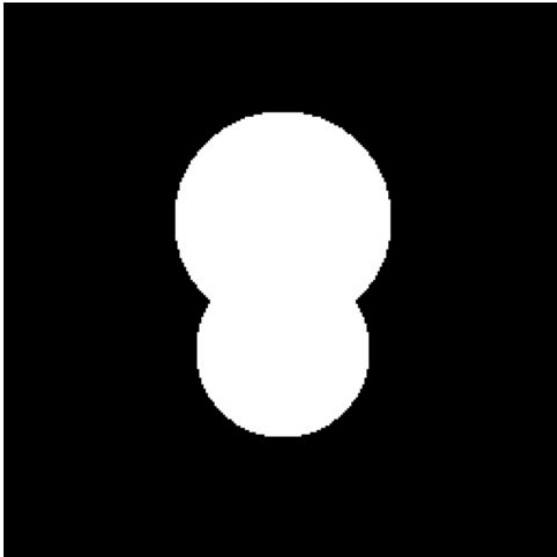
IoU: 0.0483  
Accuracy: 0.4613



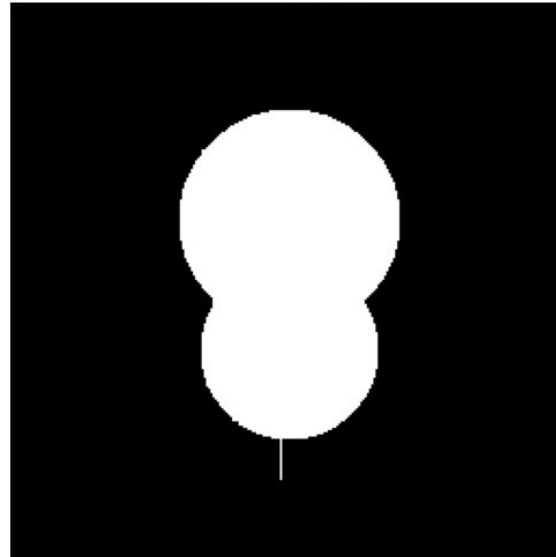
# Results

## Adding different level of noise and segmenting using the Sobel level set

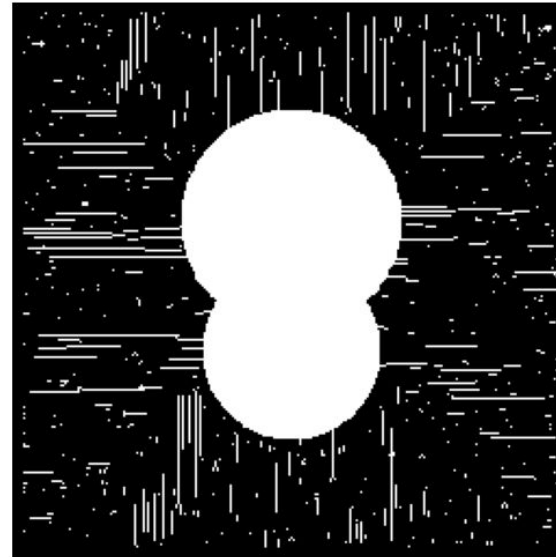
Ground Truth



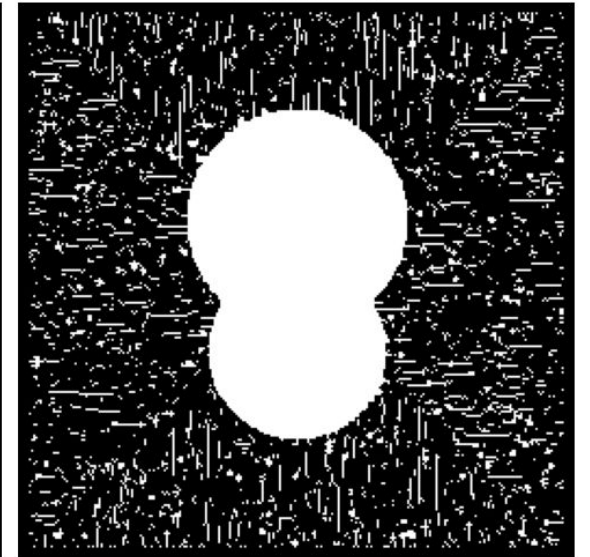
Light Noise Segmentation  
IoU: 0.9661  
Accuracy: 0.9939



Medium Noise Segmentation  
IoU: 0.7596  
Accuracy: 0.9446



Heavy Noise Segmentation  
IoU: 0.6439  
Accuracy: 0.9032





# Results

## Real World data Segmentation

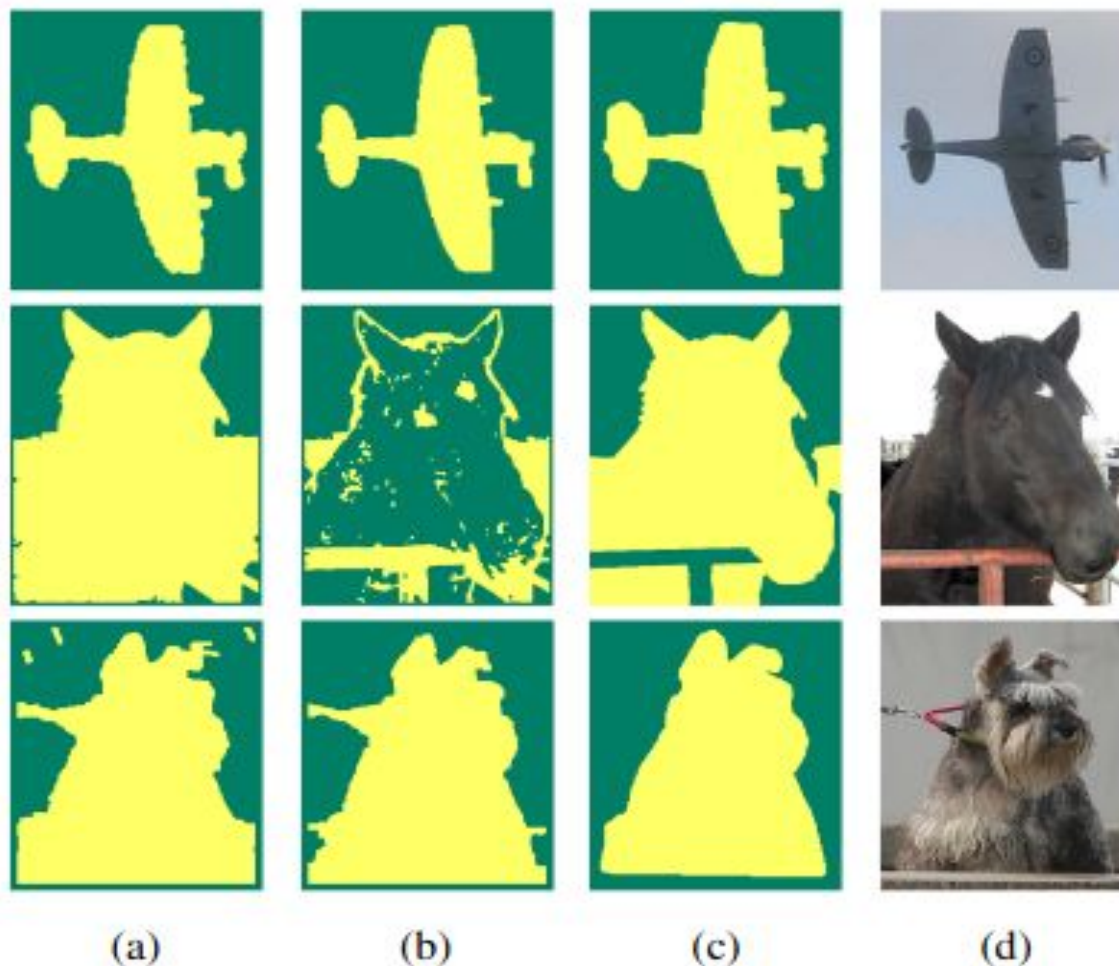


Fig. 8: Segmentation results on the Pascal VOC dataset. (a) refers to the traditional level set, (b) refers to level set with sobel filter, (c) refers to ground truth, and (d) refers to the image



# Results

Method	Light Noise	Medium Noise	Heavy Noise	Pascal VOC
LS	0.007	0.01	0.04	0.63
LS (Sobel)	0.96	0.75	0.64	0.60

## Why Sobel Level Set Struggles on Real-World Images

- **Texture Variations:** Confused by **complex** textures and smooth transitions.
- **Edge Ambiguity:** Struggles with unclear or **blended** edges.
- **Parameter Sensitivity:** Requires fine-tuned **parameters** for each scenario.