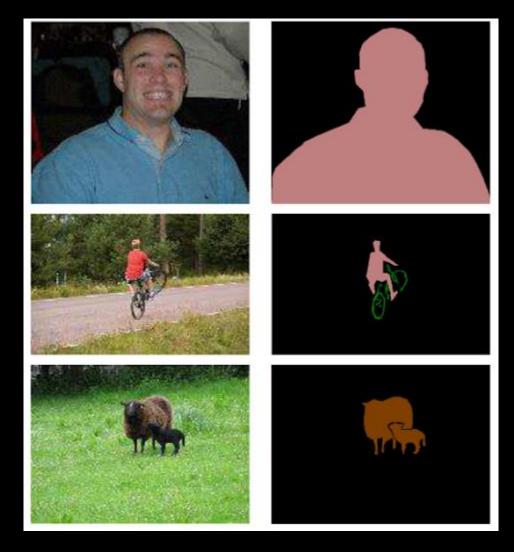


IE643 (Deep Learning: Theory & Practice) Course Project by Team – MetaNet Raj Vhora (180110058) Shivprasad Kathane (180110076)

### Outline

- Image Segmentation
- Types & Applications
- Techniques
- NN Architectures
- Loss Functions
- Medical Image Segmentation
- Pre-Processing & Training
- Challenges / Issues



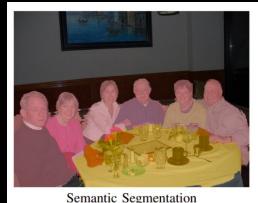
Typical Segmentation Maps [3]

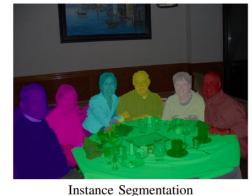
## What is Image Segmentation? [4]

- Image segmentation is a method in which a digital image is broken down into various subgroups called Image segments which helps in reducing the complexity of the image to make further processing or analysis of the image simpler
- Segmentation in easy words is assigning labels to pixels. All picture elements or pixels belonging to the same category have a common label assigned to them
- Ex: In an image containing fruit and a flower, an ideal image segmentation model would label every pixel associated with fruit as 'fruit' and similarly with every pixel associated with flower as 'flower'

### **Types** [1,2,3]

- Semantic segmentation All objects of the same type are marked with the same label
- Instance segmentation Even instances of the same type will be marked with different label
- Panoptic segmentation can be expressed as the combination of semantic & instance segmentation where each instance of an object in the image is segregated and the object's identity is predicted











(b) Semantic Segmentation





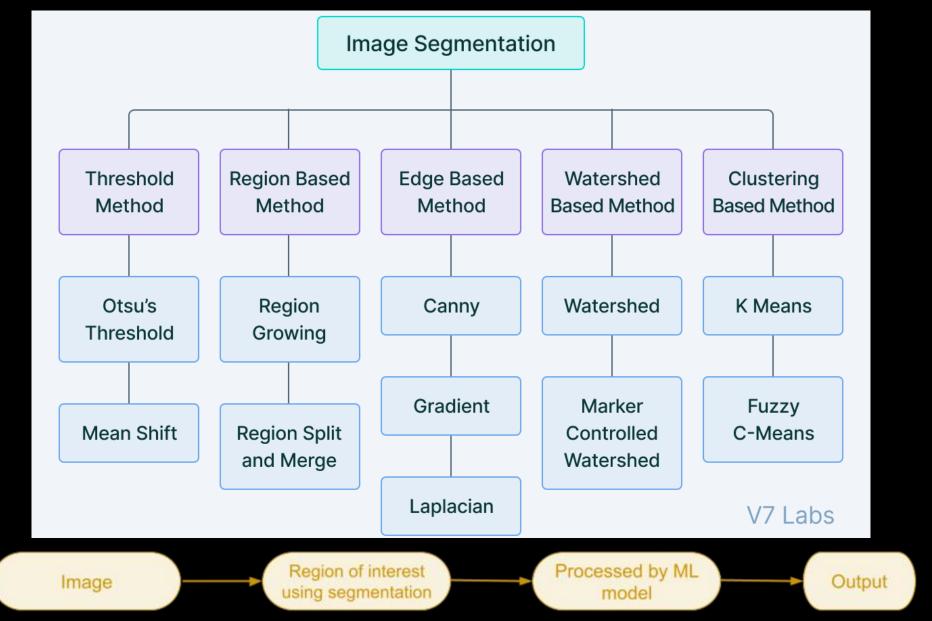
(c) Instance Segmentation

(d) Panoptic Segmentation

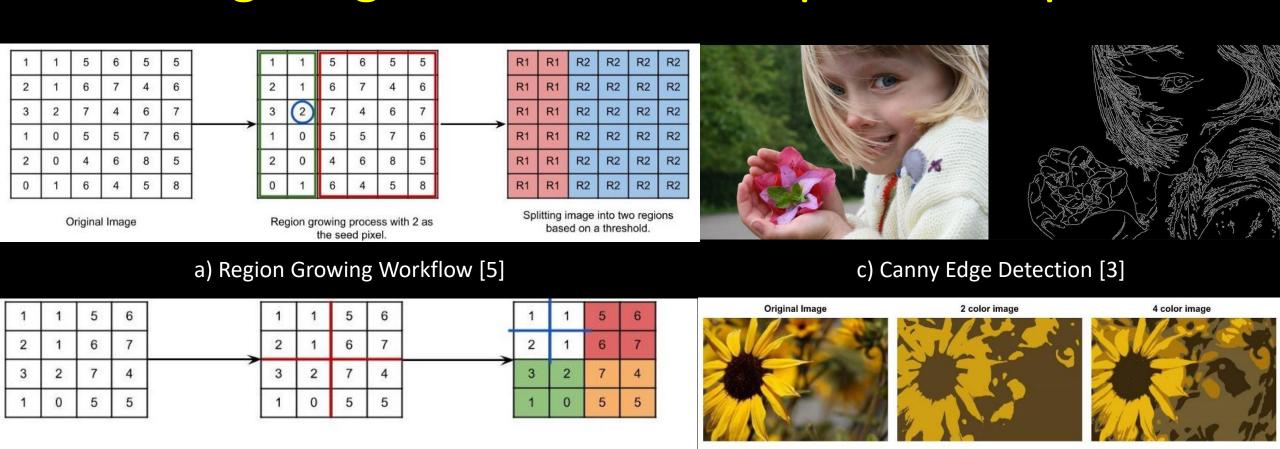
### Applications [3]

- Image segmentation finds its way in prominent fields like Robotics, Medical Imaging, Autonomous Vehicles, and Intelligent Video Analytics.
- Robotics aids machine perception and locomotion by pointing out objects in their path of motion, enabling them to change paths effectively and understand the context of their environment
- Medical Imaging helps doctors identify possible malignant features in images in a fast and accurate manner
- Smart Cities Al-based monitoring, crimes can be reported faster, road accidents can be followed up with immediate ambulances, and speeding cars can be easily caught and penalized
- Self-Driving Cars instance segmentation helps vehicles to identify road patterns and other vehicles, thereby enabling a hassle-free and smooth ride
- Image segmentation is also used by satellites on aerial imagery for segmenting out roads, buildings, and trees.

# Image Segmentation Techniques: Traditional



## Image Segmentation Techniques: Examples



6 color image

8 color image

10 color image

Classifying a quadrant as a region if it satisfies condition else performing further splitting

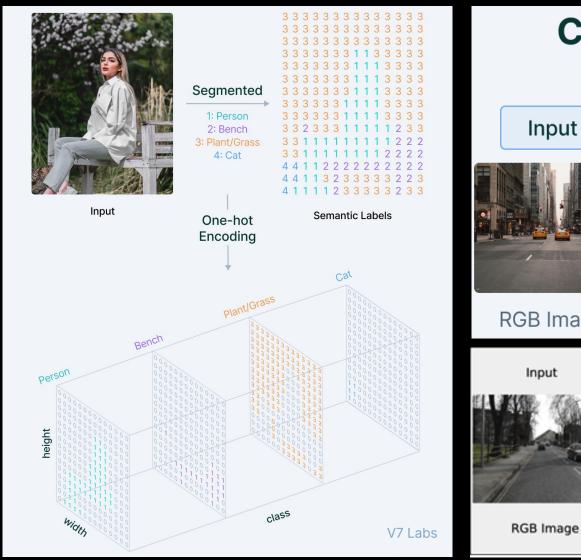
b) Region Splitting & Merging Workflow [5]

Region splitting into 4 quadrant

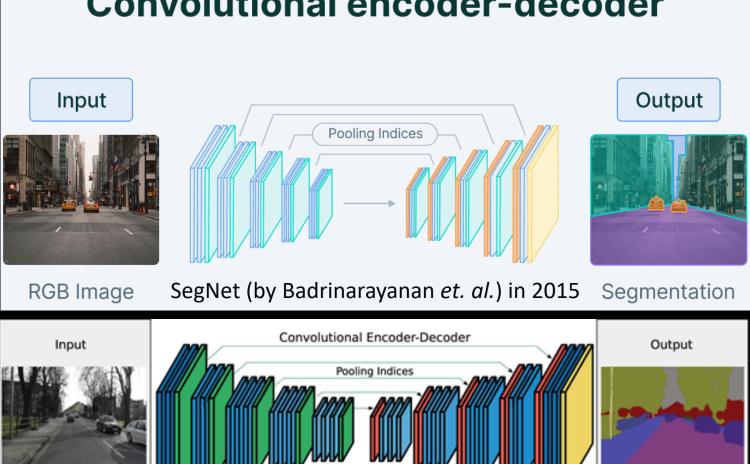
Original Image

d) Segmentation using K-Means Clustering wrt Colours [5]

# Deep Learning Based Image Segmentation



#### Convolutional encoder-decoder



Conv + Batch Normalisation + ReLU

Upsampling

Softmax

Segmentation

## Architectures: UNET, Gated SCNN, Mask RCNN, FastFCN

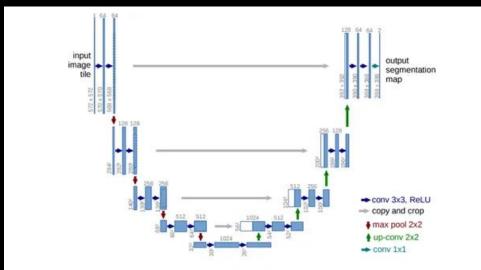


Fig. 1. U-net architecture (example for 32x32 pixels in the lowest resolution). Each blue box corresponds to a multi-channel feature map. The number of channels is denoted on top of the box. The x-y-size is provided at the lower left edge of the box. White boxes represent copied feature maps. The arrows denote the different operations.

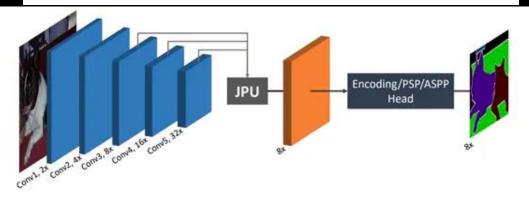


Figure 2: Framework Overview of Our Method. Our method employs the same backbone as the original FCN. After the backbone, a novel upsampling module named Joint Pyramid Upsampling (JPU) is proposed, which takes the last three feature maps as the inputs and generates a high-resolution feature map. A multi-scale/global context module is then employed to produce the final label map. Best viewed in color.

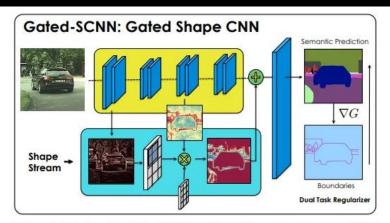


Figure 1: We introduce *Gated-SCNN* (GSCNN), a new two-stream CNN architecture for semantic segmentation that explicitly wires shape information as a separate processing stream. GSCNN uses a new gating mechanism to connect the intermediate layers. Fusion of information between streams is done at the very end through a fusion module. To predict high-quality boundaries, we exploit a new loss function that encourages the predicted semantic segmentation masks to align with ground-truth boundaries.

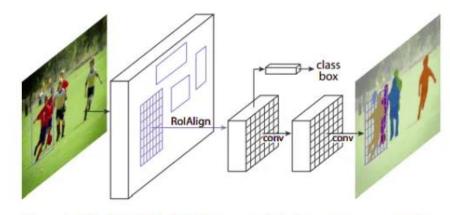
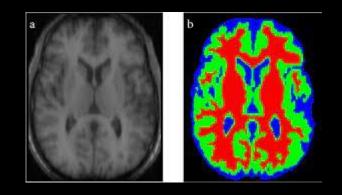


Figure 1. The Mask R-CNN framework for instance segmentation.

### **Loss Functions**

- Focal Loss  $FL(p_t) = -(1 p_t)^{\gamma} \log(p_t)$ .
  - Improvement to standard cross entropy loss
  - This is done by changing its shape such that the loss assigned to well-classified examples is down-weighted
- Dice Loss  $DSC = \frac{2|X \cap Y|}{|X| + |Y|}$
- Intersection over Union (IoU) balanced loss  $I_{OU = TP/(TP + FP + FN)}$
- Boundary Loss  $\text{Dist}(\partial G, \partial S) = \int_{\partial G} \|y_{\partial S}(p) p\|^2 dp$
- Weighted Cross Entropy  $WCE(p, \hat{p}) = -(\beta p \log(\hat{p}) + (1-p) \log(1-\hat{p}))$
- Lovász-Softmax loss  $loss(f) = \frac{1}{|C|} \sum_{c \in C} \overline{\Delta_{J_c}}(m(c))$

## Medical Image Segmentation

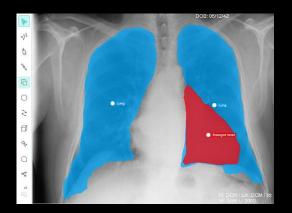


- Medical Imaging is an important domain of computer vision that focuses on the diagnosis of diseases from simple visual data & biomedical scans.
- Medical image segmentation involves the extraction of regions of interest (ROIs) from 3D image data, such as from Magnetic Resonance Imaging (MRI) or Computed Tomography (CT) scans.
- The main goal is to identify areas of the anatomy required for a particular study, for example, to simulate physical properties or virtually positioning CAD-designed implants within a patient. It mainly helps doctors identify possible malignant features in images in a fast and accurate manner.
- Medical image segmentation can be a time-consuming task, and recent advances in AI are making it easier for routine tasks to be completed.

[3, 6]

# **Pre-Processing & Training**

- Data Annotation for Labelling of Images
- Data Augmentation Techniques like:
  - a) Identity transformation
- b) Rotation by random angle
- c) Flipping along random axis
- d) Generating mirror images
- Image Processing Techniques like:
  - Downsampling = To reduce pixels for easy memory handling
  - Cropping = To get same sizes with principal region of focus
- Network Training Techniques like:
  - a) Deeply Supervised b) Weakly Supervised c) Transfer Learning



## Challenges / Issues

- Limited Annotated Data
- Lack of Healthy Brain Images
- Class Imbalance
- Overfitting & Long Training Times
- Vanishing Gradients
- Heterogeneous Nature of Visible Organs
- 3D Images → Computationally Expensive

### References

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- 3. <a href="https://www.v7labs.com/blog/image-segmentation-guide">https://www.v7labs.com/blog/image-segmentation-guide</a>
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- 6. https://www.synopsys.com/glossary/what-is-medical-image-segmentation.html
- 7. M.H. Hesamian, W. Jia, X. He et al., "Deep Learning Techniques for Medical Image Segmentation: Achievements and Challenges", J Digit Imaging 32, 582–596 (2019) <a href="https://doi.org/10.1007/s10278-019-00227-x">https://doi.org/10.1007/s10278-019-00227-x</a>
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