

(11)

# Detection & Segmentation & Localization

Semantic Segmentation

Classification  
+ Localization

Object Detection

Instance Segmentation

## ★ Semantic Segmentation

Image (Input) → decision of category for every Pixel in the image (Output)

⇒ Don't differentiate instances, only care about pixels

### Idea 1: Sliding Window

↳ Problem: Very inefficient! Not reusing shared features between overlapping patches.

### Idea 2: Fully Convolutional network

⇒ Design a network as a bunch of convolutional layers to make predictions for pixels all at once.

⇒ Design network as a bunch of convolutional layers, with downsampling & upsampling inside the network.

### # nn-Netwrok upsampling: Unpooling

1	2
3	4

1	1	2	2
1	1	2	2
3	3	4	4
3	3	4	4

1	2
3	4

Bed of Nails

1	0	2	0
0	0	0	0
3	0	4	0
0	0	0	0

Nearest Neighbor



## Max Pooling

1	2	6	3
3	5	2	1
1	2	2	1
7	3	4	8



5	6
7	8

Rest of Network

1	2
3	4

## Max Unpooling

0	0	2	0
0	1	0	0
0	0	0	0
3	0	0	4

Remember which element was max!

Use position from Pooling layer.

## # Learnable Upsampling: Transpose Convolution


2x2




4x4

⇒ It is a filter for upsampling.

## \* Classification + Localization



FC layer → 1000

Class Scores → Softmax Loss

Correct Label



4096 → 4

Box Coordinate (x, y, w, h)

→ L2 Loss

Correct box

⇒ Treat localization as a regression problem!



## \* Object Detection

- ⇒ Given an image draw bounding box around all the class in the image and give the class label.
- ⇒ Each image needs a different number of output.
- ⇒ One approach: Sliding Window
  - ↳ Problem: Need to apply CNN to huge number of locations & scales, very computationally expensive!

## # Region Proposal

- ⇒ Find "blobby" image regions that are likely to contain objects.

(R-CNN)

{Region Based Convolutional Neural Networks}

(Fast R-CNN)

(Faster R-CNN)

↳ Jointly train with 4 losses:

- ① → RPN classifies object/not object
- ② → RPN suggests box coordinate
- ③ → Find classification score (object class)
- ④ → Find box coordinate



## ⊕ Detection without Proposals: YOLO/SSD

{ You only Look once }

{ Single-Shot Detection }

⇒ Within each grid cell:

→ Regress from each of the  $B$  base boxes to a final box with 5 numbers.

( $dx$   $dy$   $dh$   $dw$ , confidence)

→ Predict score for each of  $C$  classes.  
(Including background as a class)

## ★ Instance Segmentation

Mask R-CNN

