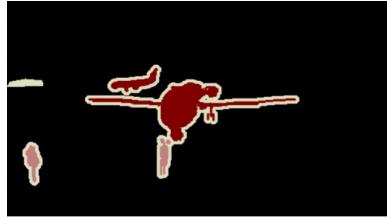


Semantic Segmentation:



← 주어진 이미지의 모든 픽셀에 대해 classification을 진행

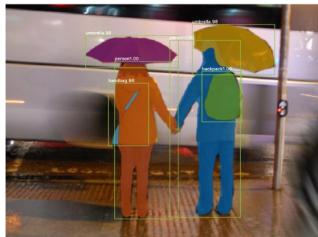
[**경계**: background (foreground가 아닌 부분)]

[**붉은색**: 배경부]

[**분홍색**: 사람들]

{ 클래스가 같은 픽셀은 모두 같은 label을 갖고 있음
scene의 전체에 대한 이해가 필요할 때 유용
ex.) bio 분야에서 세포 등을 관찰

Instance Segmentation:



- 모든 픽셀에 대해 추론하지 않음
(오직 foreground를 찾아내는 것에 집중)
- 각 객체마다 찾아낸 bounding box를 기준으로 하여 그 안의 한 가지 object에 대해서 segmentation 작업을 진행한다.
- 결과적으로, object에 해당하는 픽셀에 대해서만 binary classification을 함.
- 같은 class에 해당하는 instance 하나하나를 구별함.

{
픽셀이 (class, id)의 pair을 추운

→ id는 같은 class의 서로 다른 instance를 구분하는데

실시간으로 물체를 찾아내는 작업 같이 속도가 중요시되는 일에 유용

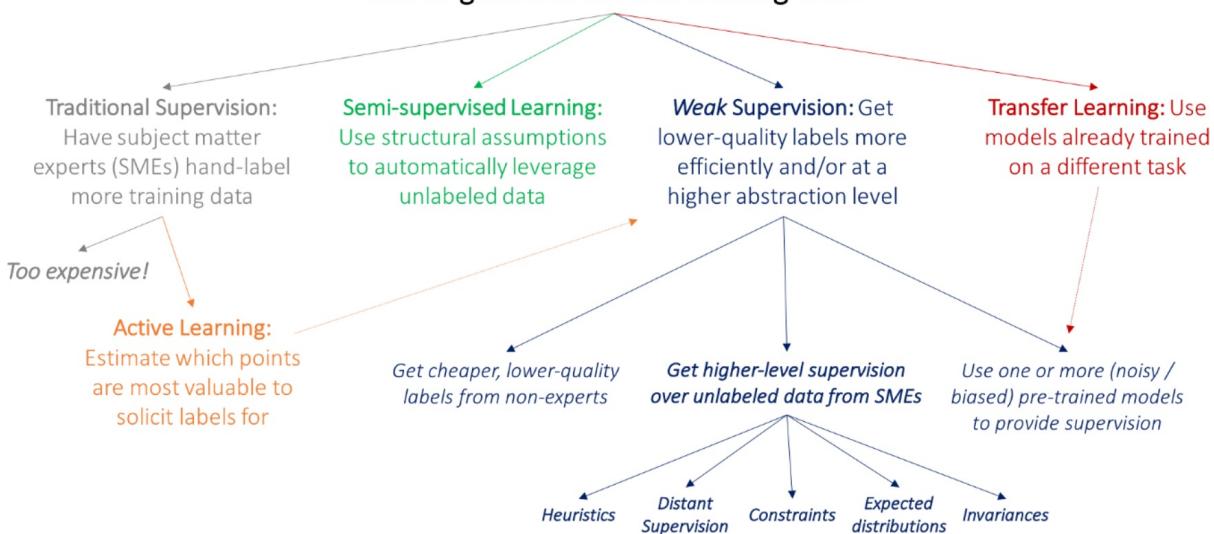
< Weakly Supervised Learning >

When to use:

- The available labeled data is insufficient to obtain a supervised model with good performance.
- The available labeled data is noisy or obtained from an imprecise source.
- There are insufficient domain experts, or getting them is very costly.
- The available time for manual annotation is very limited.

→ Weakly supervised learning is to decrease the cost of human experts annotation and increase the available labeled data for training.

How to get more labeled training data?



Near Supervision

- ① Incomplete supervision: only a subset of training data are labelled
- ② Inexact supervision: the training data are given with labels but not as exact as desired
- ③ Inaccurate supervision: in the training data, there are some labels with mistakes.

<Incomplete supervision>

- active learning: domain expert
- semi-supervised: generative models, label propagation, TSVM

CAM : bounding box이란 것도, CAM을 사용하면 bounding box가 위치한 곳을 가리킬 수 있으므로 labeling 할 때, bounding box를 정해주는 도구로써 사용할 수 있다.

global max pooling: ordinary maxpooling layer with pool size equals to the size of the input.

$$\text{ex) } \begin{matrix} 0, 1, 2, 2, 5, 1, 2 \\ \text{input} \end{matrix} \rightarrow \begin{matrix} 5 \\ \text{output} \end{matrix}$$

[Weakly Supervised Learning of Instance Segmentation with Inter-pixel Relations]

↑ Inexact supervision the training data are given labels but not as exact as desired.

: Generates pseudo instance segmentation labels of training images

{
X distinguish different instances (instance-agnostic)
X find entire instance areas with accurate boundaries.

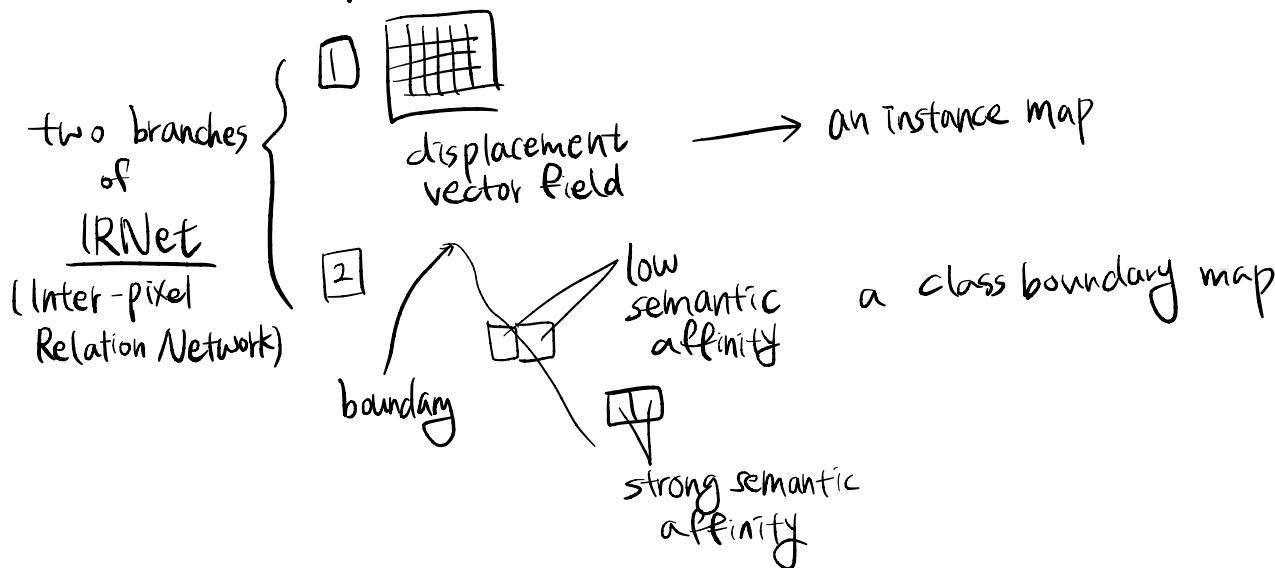
Complement CAM

→ a class-agnostic instance map

→ pairwise semantic affinity



Generate a psuedo instance segmentation label by selecting the instance label with the highest attention score in the instance-wise CAMS at each pixel.



↓
Collect pixels with high attention scores

& train IRNet with the displacements and class equivalence between the collected pixels.

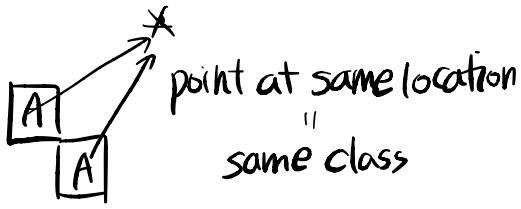
∴ No supervision in addition to image-level class labels is required.

→ weak supervised learning

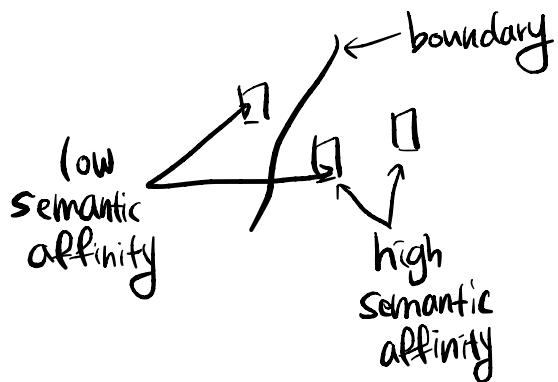
Displacement vector field: the centroid of the instance the pixel belongs to

0

Convert
Instance map



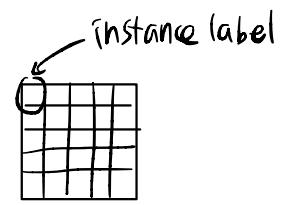
Class Boundaries : compute semantic affinities between pairs



< Roles of Class Attention Map (CAM) >

① define seed areas of instances $\dots \rightarrow$ entire instance areas.
reliable later

② extract/infer pixel relations $\dots \rightarrow$ source of supervision for I²Net



Displacement $D \rightarrow$ class-agnostic instance map (I)

: a rough instance segmentation mask without class labels nor accurate boundaries.

Class Boundary $B \rightarrow$ pairwise affinities

\downarrow
semantic
affinities

Confidence score for
class equivalence between
a pair of pixels.

incorporation of
instance-agnostic CAMs \rightarrow instance-wise
CAMs