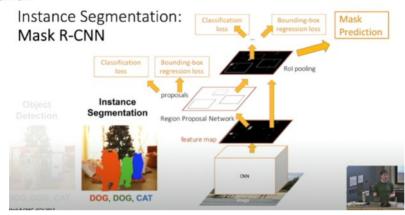
6 Segmentation

2024年12月18日

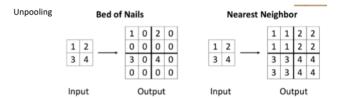
label each pixel in the image with a label. 不知道哪几个pixel属于cow, 有多少只cow

- 1. patching window. inefficient, not shared features between patches
- 2. fully conv: bunch of conv o make predictions for pixels all at once.
 - a. loss: per-pixel cross-entropy
 - b. all conv
 - i. effective receptive field size is linear.
 - $i\,i$. expensive computation
 - c. conv downsampling then upsampling
- 3. Things (individual obj instances), stuff (cannot separa)
 - a. obj detection: only things
 - b. segmentation: semantic:both; instance segmentation (both, obj detec then predict a seg)
- 4. Mask R-CNN



fix functions:

C x 2 x 2

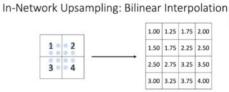


 $C \times 4 \times 4$

Not and the second seco

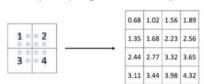
C x 2 x 2

C x 4 x 4



 $\begin{aligned} & \text{Input: C x 2 x 2} & \text{Output: C x 4 x 4} \\ f_{x,y} &= \sum_{i,j} f_{i,j} \max(0,1-|x-i|) \max(0,1-|y-j|) & i \in \{\lfloor x \rfloor -1,\dots,\lceil x \rceil +1\} \\ & \text{Use two closest neighbors in x and y} & j \in \{\lfloor y \rfloor -1,\dots,\lceil y \rceil +1\} \end{aligned}$

In-Network Upsampling: Bicubic Interpolation



Input: C x 2 x 2 Output: C x 4 x 4

Use **three** closest neighbors in x and y to construct **cubic** approximations (This is how we normally resize images!)

和前一步的上采样结合

Learnable upsampling: transposed convolution