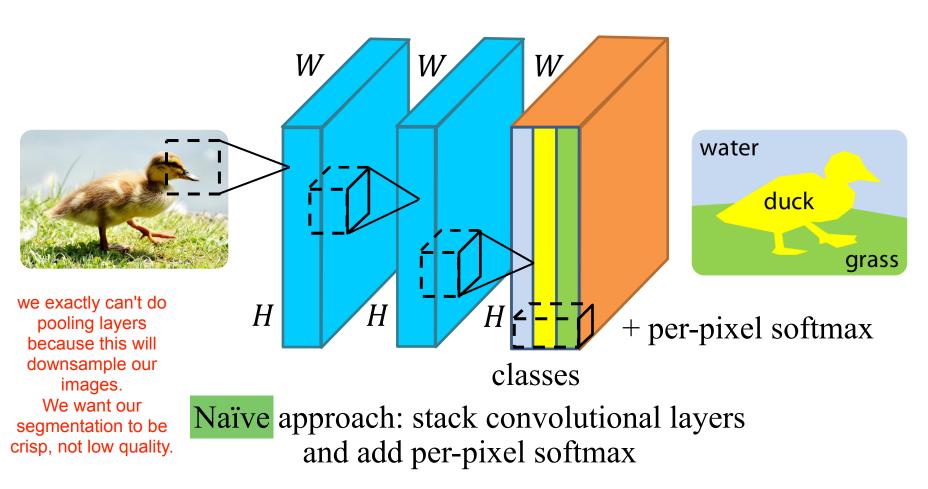
We need to classify each pixel



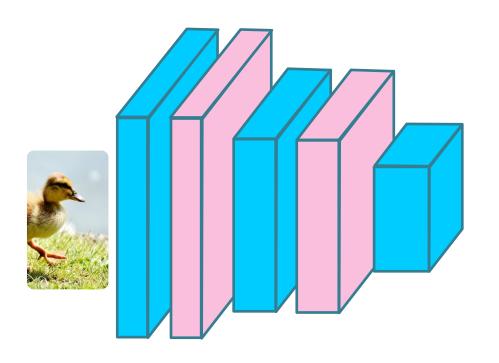
image classification + localization. Not only what is the image, but where we see it.

We need to classify each pixel



We go deep but don't add pooling, too expensive

Let's add pooling, which acts like down-sampling

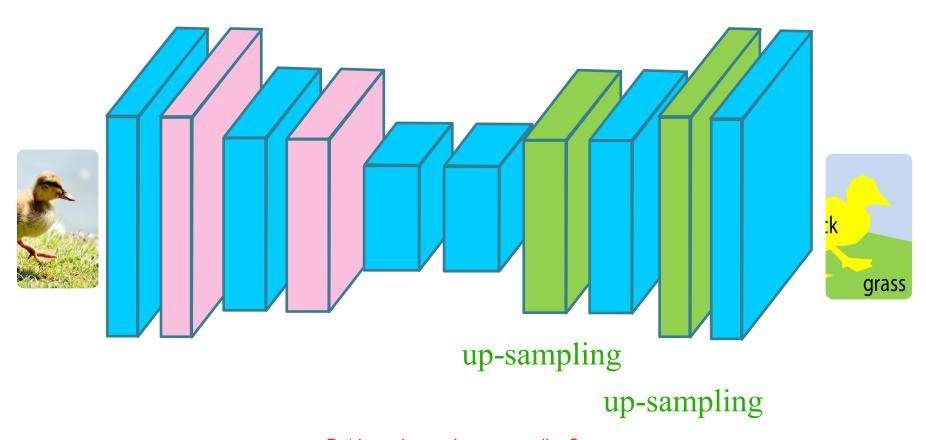


Wait a second!
We need to classify each pixel!

Need to do unpooling!

Yes. so we do the 'reverse'. Upsampling.

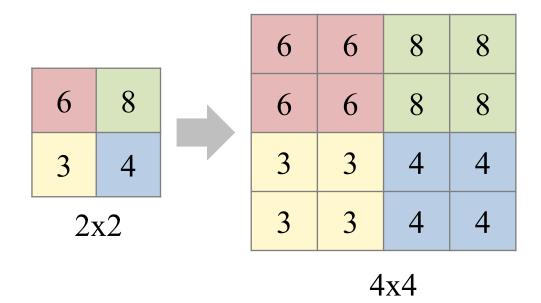
Let's add pooling, which acts like down-sampling



But how do we do up-sampling?

# Nearest neighbor unpooling

Fill with nearest neighbor values

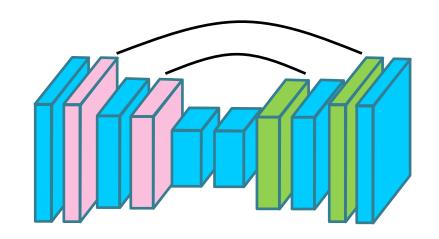


Pixelated and not crisp!

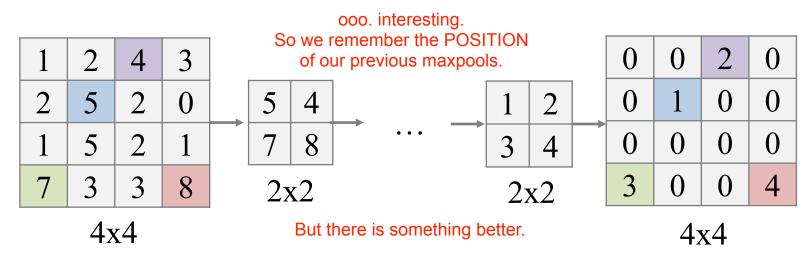
Can't just use this.

## Max unpooling

Corresponding pairs of downsampling and upsampling layers

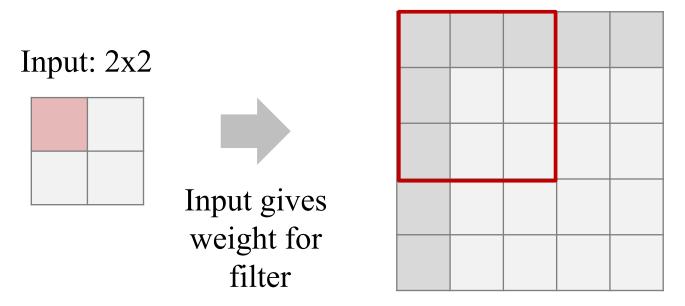


Remember which element was max during pooling, and fill that position during unpooling:



# Learnable unpooling

- Previous approaches are not data-driven!
- We can replace max pooling layer with convolutional layer that has a bigger stride!
- What if we can apply convolutions to do unpooling?

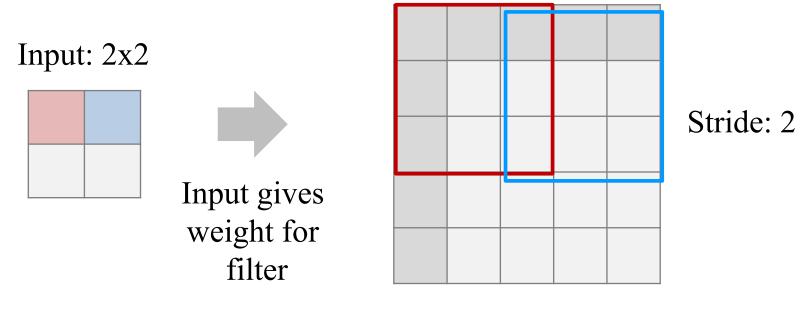


Output: 4x4

The step is to take the value from the red input cell, and then multiply it with the kernel's values to give the output in the red outlined box in the output 4x4.

# Learnable unpooling

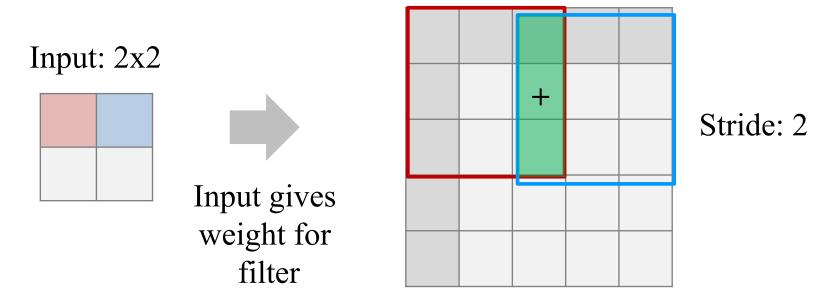
- Previous approaches are not data-driven!
- We can replace max pooling layer with convolutional layer that has a bigger stride!
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Output: 4x4

# Learnable unpooling

- Previous approaches are not data-driven!
- We can replace max pooling layer with convolutional layer that has a bigger stride!
- What if we can apply convolutions to do unpooling?

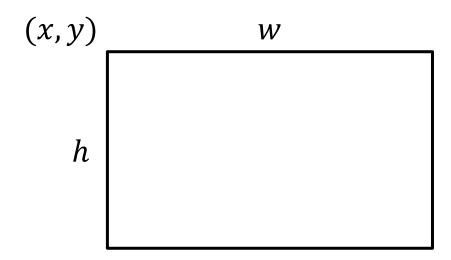


For overlaps, just take sum.

Output: 4x4

#### **Object classification + localization**

We need to find a bounding box to localize an object.

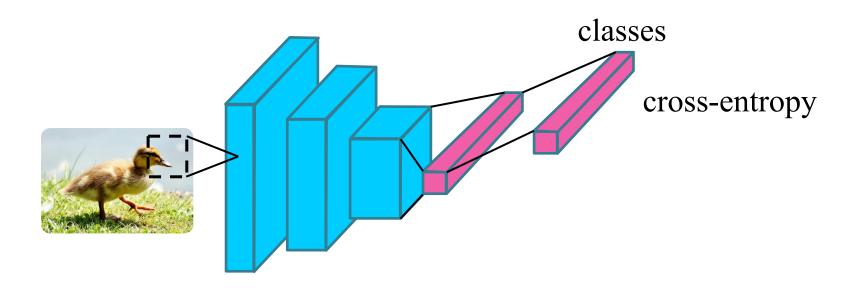


We will use regression for (x, y, w, h)!

(x,y) is the coordinate of the upper left corner of box. w,h is width and height.

### **Object classification + localization**

#### Classification network:

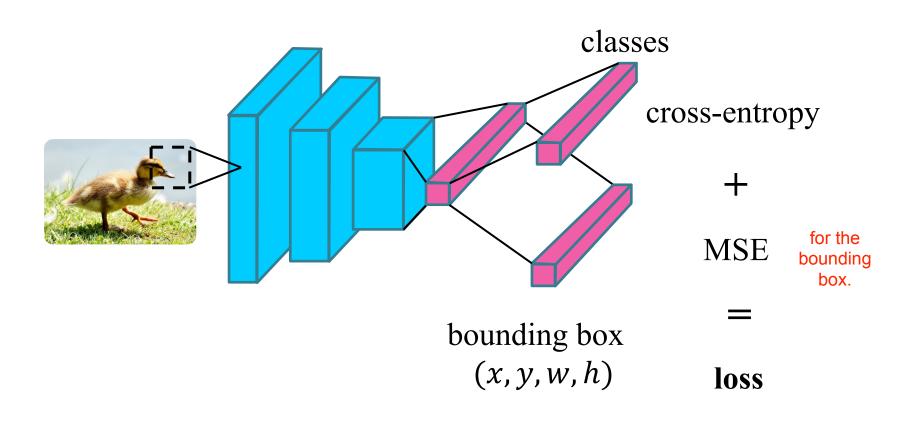


Do we need a second network?

No! Just use the same feature extractors, and train a smaller network on top of it.

#### **Object classification + localization**

Classification + localization network:



### **Summary**

- In this video we took a sneak peek into other computer vision problems that successfully utilize convolutional neural networks.
- This video concludes our introduction to neural networks for images!