

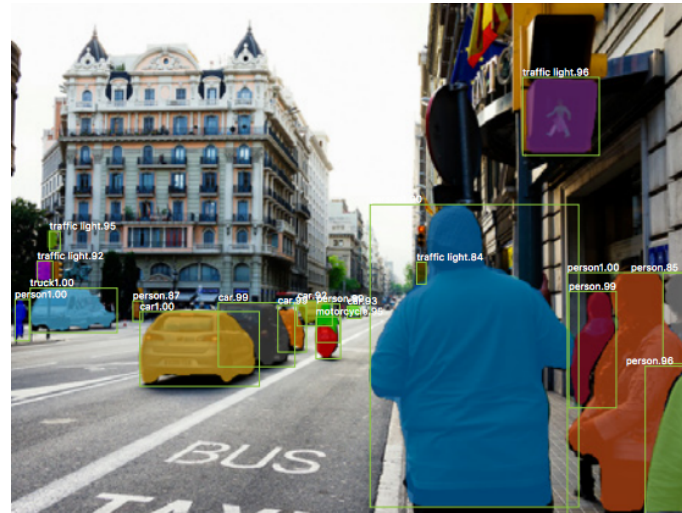
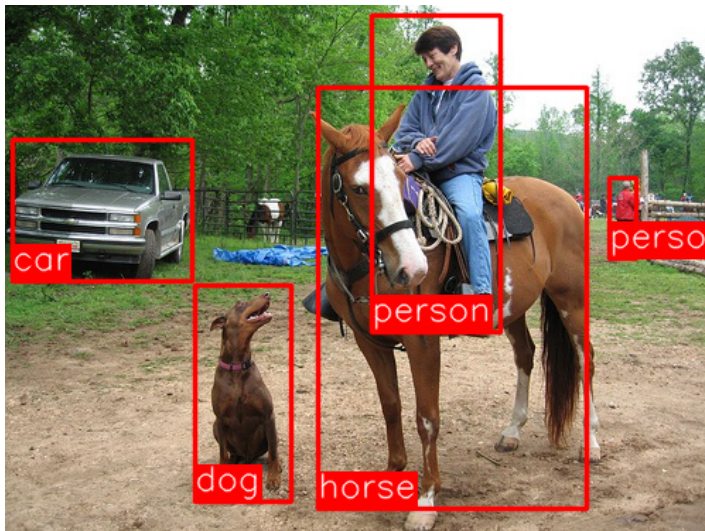
# Object detection and segmentation

ISTD 50.035

Computer Vision

# Object detection / segmentation

- Finding different objects in an image and classify them

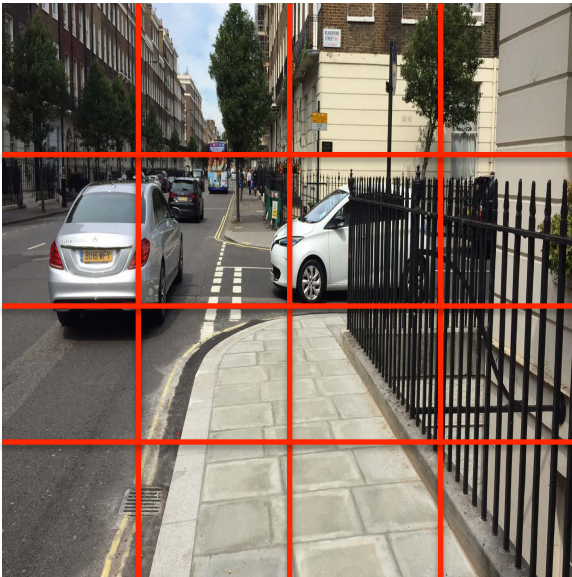


# YOLO: You Only Look Once

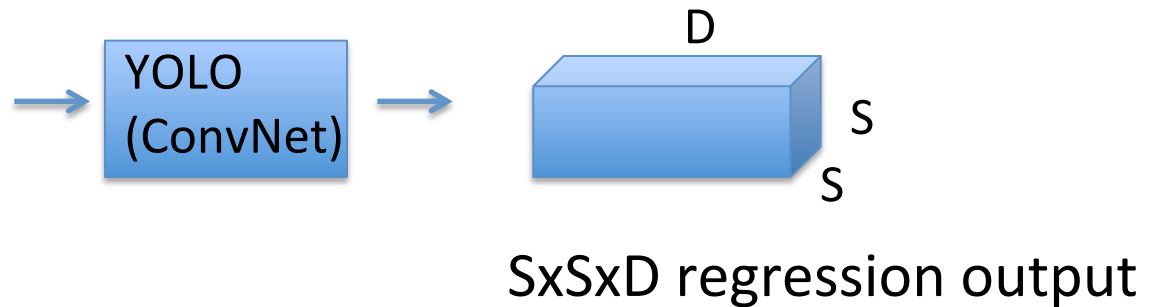
- Reframe object detection as a single regression problem
- Very fast: object class probabilities and bounding box coordinates regression in a single forward pass

# YOLO: You Only Look Once

SxS grid



Each D-dim vector encodes the class probabilities and bounding box coordinates for that cell

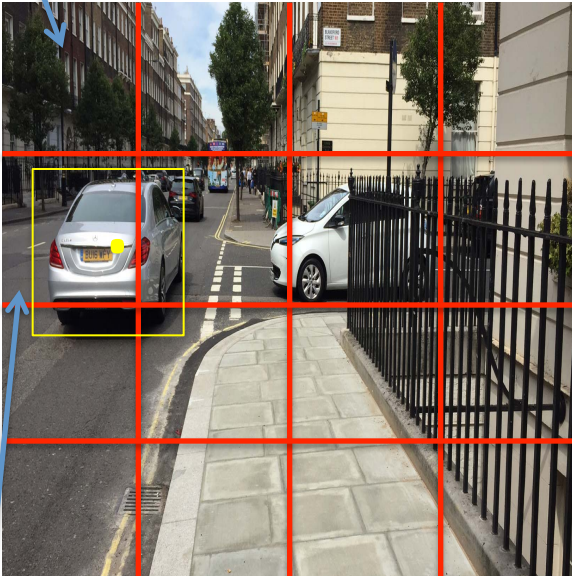


- box confidence (object or not, how accurate is this box)
- x,y,h,w (center coord, width, height)
- class probabilities

# YOLO: You Only Look Once

[0,\_,\_,\_,\_,\_]

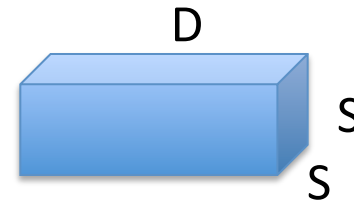
SxS grid



[1,0.9,0.6,0.25,0.25,0,1]

Dog, Car

Each D-dim vector encodes the class probabilities and bounding box coordinates for that cell



SxSxD regression output

-box confidence (object or not, how accurate is this box)

-x,y,h,w (center coord, width, height)

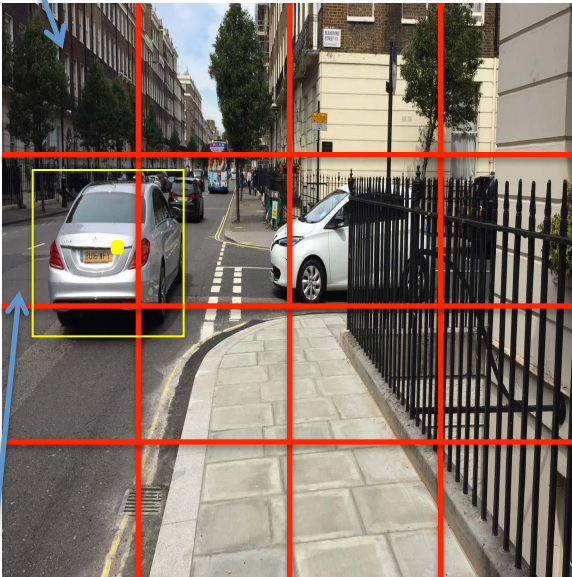
-class probabilities

- If the center of an object falls into a grid cell, that grid cell is responsible for detecting that object
- Each grid cell detects only one object (small cell is used)

# YOLO: You Only Look Once

[0,\_,\_,\_,\_,\_]

SxS grid



[1,0.9,0.6,0.25,0.25,0,1]

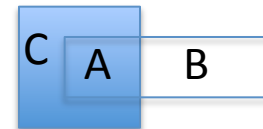
Dog, Car

Center (x,y): relative to the grid,  
normalized to [0,1]

(w,h): relative to image size, [0,1]

Box confidence:  $\text{Pr}(\text{object}) * \text{IOU}(\text{pred}, \text{truth})$

- No object -> box confidence = 0
- Higher IOU -> potentially more accurate bbox



Intersection over  
Union:

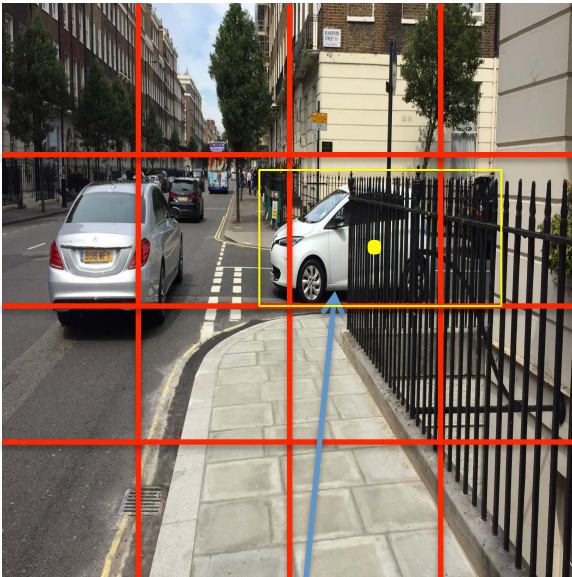
$\text{IOU} = A / (A+B+C)$   
Usually requires  $> 0.5$   
for overlapping

-If the center of an object falls into a grid cell, that grid cell is responsible for detecting that object  
-Each grid cell detects only one object (small cell is used)



# YOLO: You Only Look Once

SxS grid

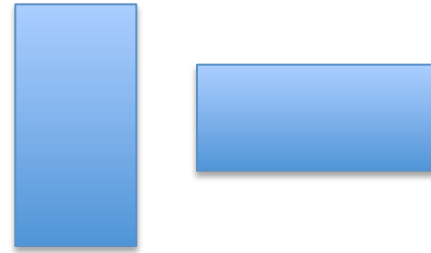


[0,\_,\_,\_,\_,1,0.6,0.6,0.24,0.3,0,1]

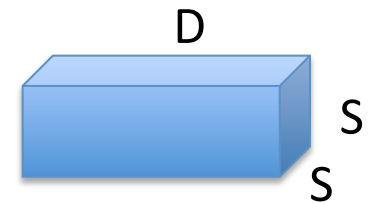
bbox 0

bbox 1

Predict two bounding boxes (anchors)  
per grid cell



Training: only want one bbox predictor to be  
responsible for each object: one with highest IOU

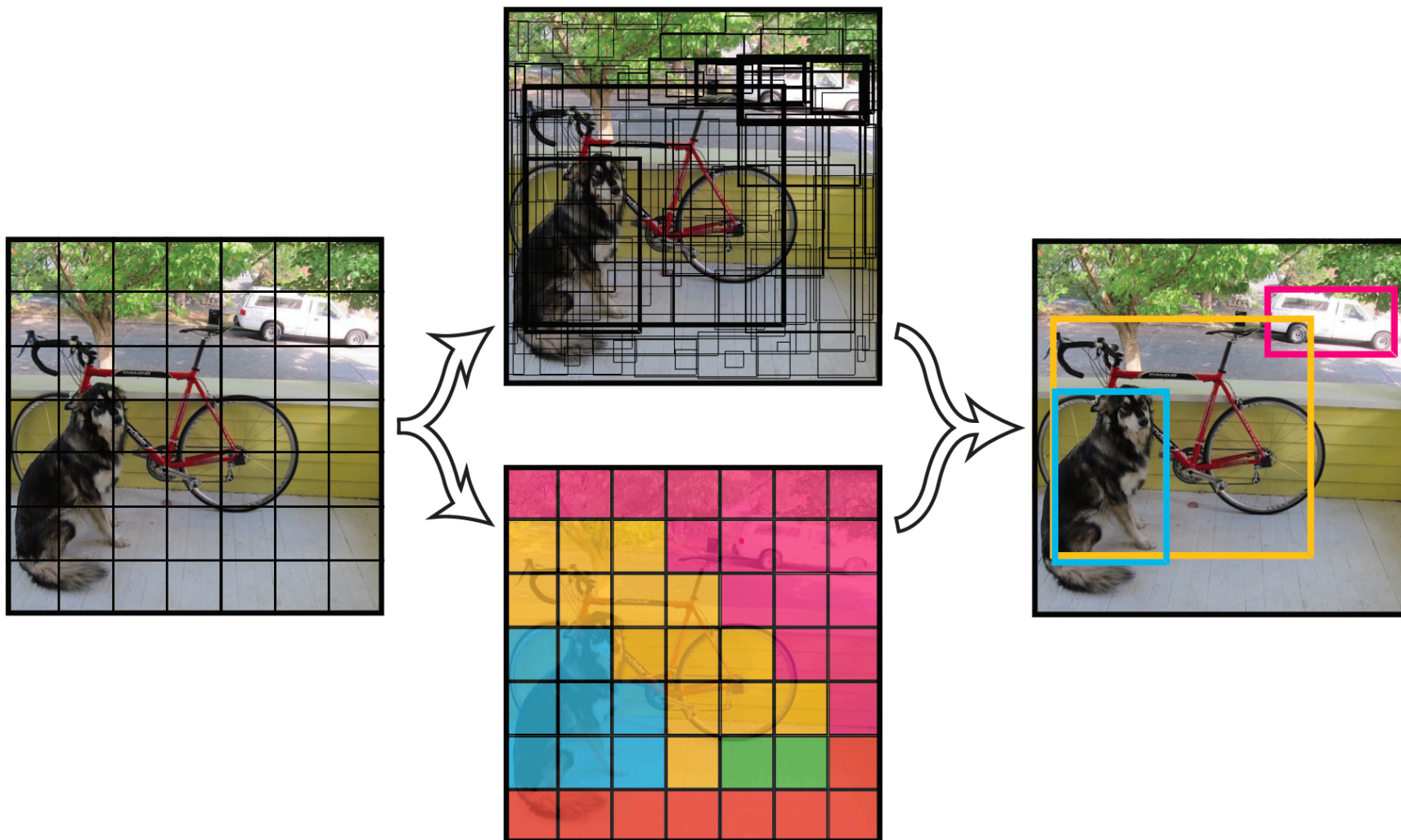


YOLO output: SxSxD tensor

$D = B \times 5 + C$

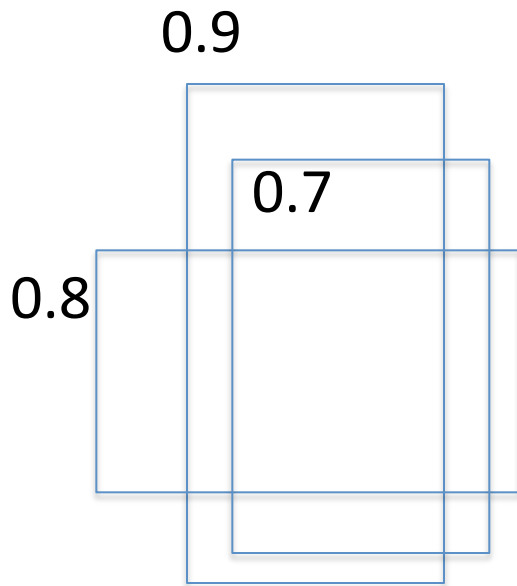
$B = 2$  (two bbox)

# YOLO: You Only Look Once





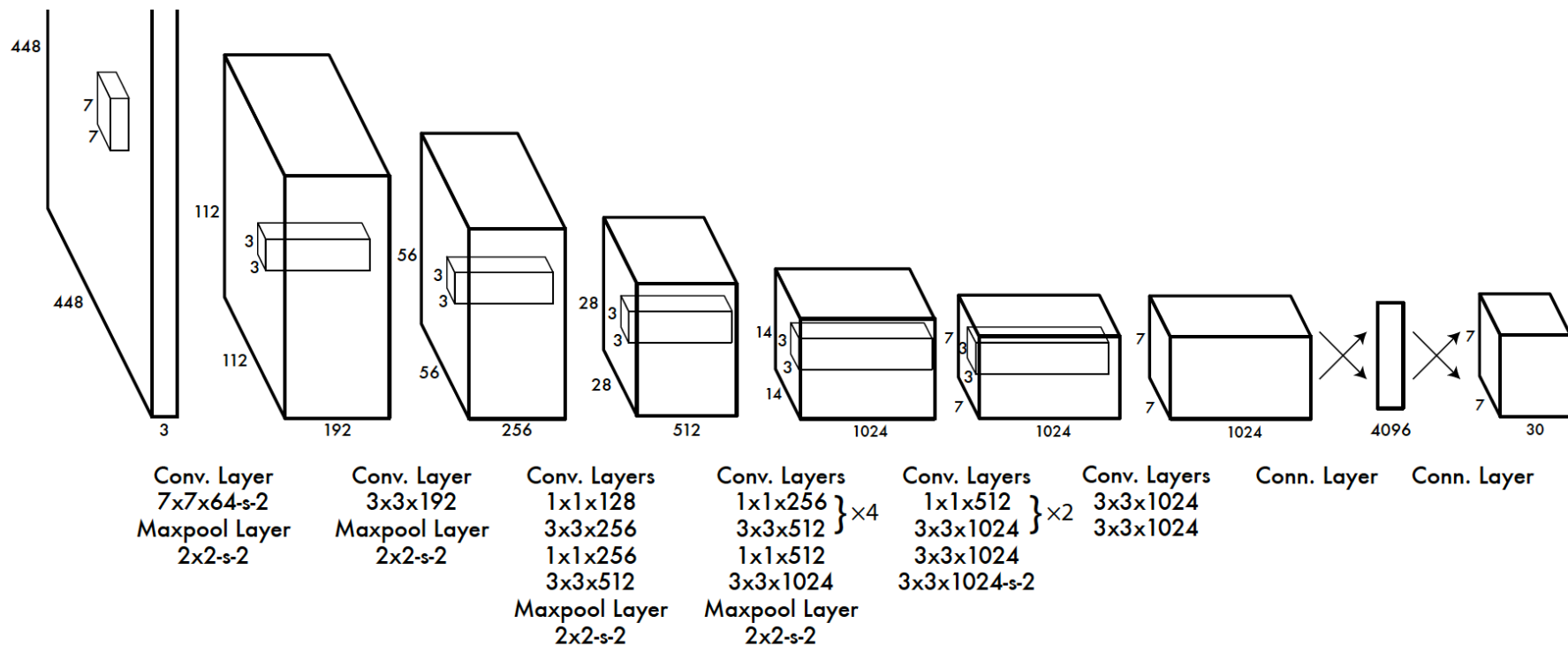
# Non maximum suppression



Many detection per object -> choose one

- 1) Discard bbox\_confidence  $\leq 0.6$
- 2) Select one with highest bbox\_confidence
- 3) Discard bbox with IOU  $\geq 0.5$
- 4) Goto (2)

# YOLO: You Only Look Once



20 labeled classes

# YOLO: You Only Look Once

Training: multi part loss

$$\begin{aligned} & \lambda_{\text{coord}} \sum_{i=0}^{S^2} \sum_{j=0}^B \mathbb{1}_{ij}^{\text{obj}} (x_i - \hat{x}_i)^2 + (y_i - \hat{y}_i)^2 \\ & + \lambda_{\text{coord}} \sum_{i=0}^{S^2} \sum_{j=0}^B \mathbb{1}_{ij}^{\text{obj}} \left( \sqrt{w_i} - \sqrt{\hat{w}_i} \right)^2 + \left( \sqrt{h_i} - \sqrt{\hat{h}_i} \right)^2 \\ & + \sum_{i=0}^{S^2} \sum_{j=0}^B \mathbb{1}_{ij}^{\text{obj}} \left( C_i - \hat{C}_i \right)^2 \\ & + \lambda_{\text{noobj}} \sum_{i=0}^{S^2} \sum_{j=0}^B \mathbb{1}_{ij}^{\text{noobj}} \left( C_i - \hat{C}_i \right)^2 \end{aligned}$$

where  $\mathbb{1}_i^{\text{obj}}$  denotes if object appears in cell  $i$  and  $\mathbb{1}_{ij}^{\text{obj}}$  denotes that the  $j$ th bounding box predictor in cell  $i$  is “responsible” for that prediction.

$$+ \sum_{i=0}^{S^2} \mathbb{1}_i^{\text{obj}} \sum_{c \in \text{classes}} (p_i(c) - \hat{p}_i(c))^2$$