

YOLO

(You Only Look Once)

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INTRODUCTION

- Object detection is a computer technology related to computer vision and Image processing
- Fast, accurate algorithms for object detection would allow computers to drive cars without specialized sensors, enable assistive devices to convey real time scene information to human users.
- Current detection systems repurpose classifiers to perform detection means the systems take a classifier to detect object and evaluate it at various locations and scales in a test image.
- You Only Look Once (YOLO), a new approach to object detection.

Previously : Object Detection by Classifiers

- **DPM (Deformable Parts Model)**

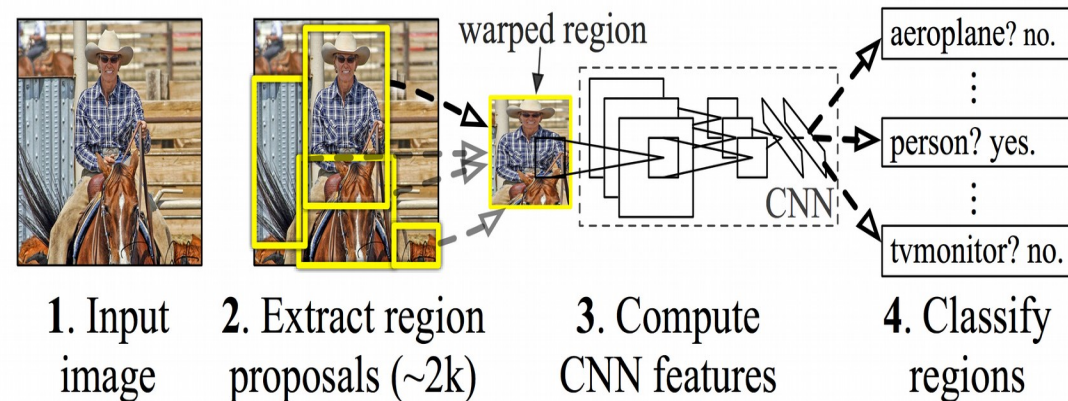
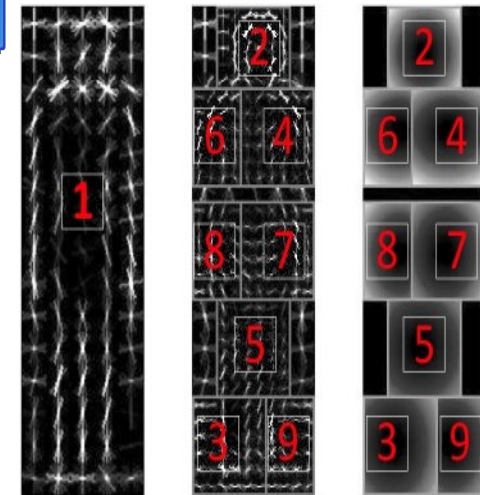
- ◆ use a sliding window approach where the classifier is run at evenly spaced locations over the entire image.

- **R-CNN**

- ◆ use regional proposal methods to first generate potential bounding boxes
- ◆ run a classifier on the proposed boxes
- ◆ Post processing

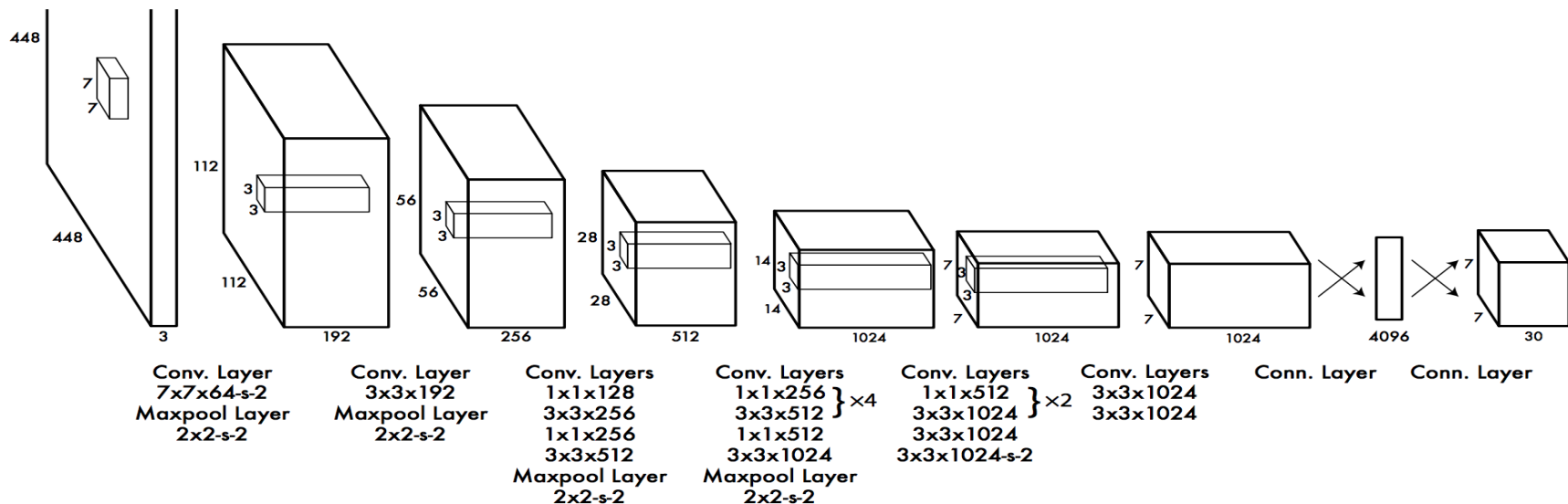
- **YOLO**

- ◆ Resize image, run convolutional network
- ◆ non-max suppression.



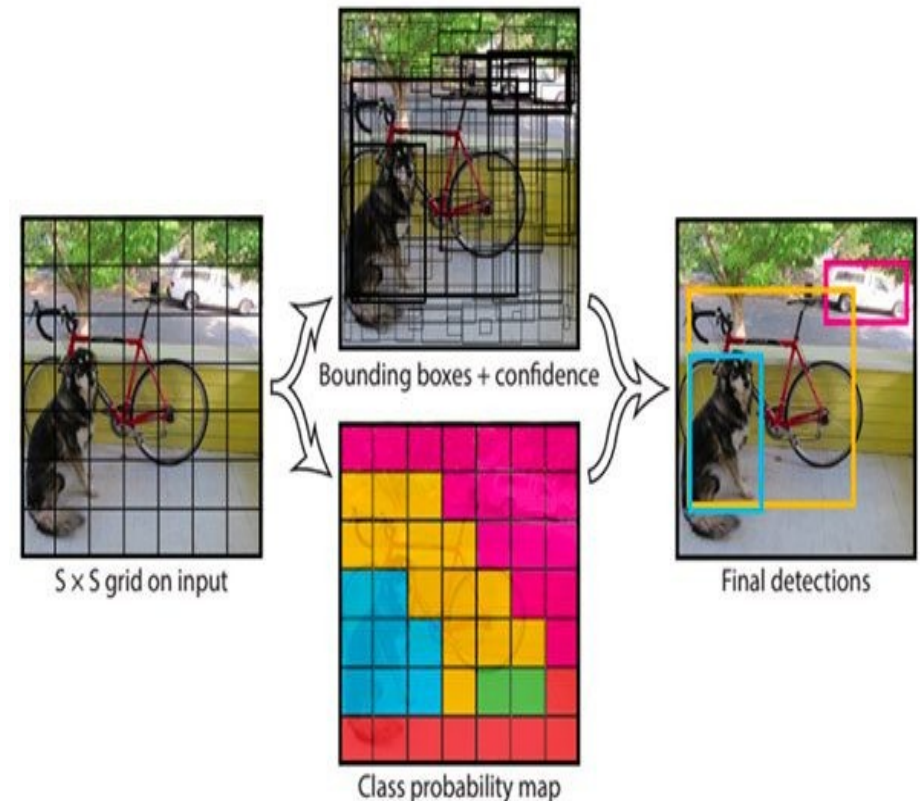
YOLO

- A single regression Problem (single neural network), straight from image pixels to bounding box coordinates and class probabilities
- Predicts directly from full images in one evaluation, about all classes
- The Architecture : Inspired by the GoogleNet
- 24 convolutional layers followed by the 2 fully connected layers.



The Model

- This models detection as a regression problem.
- It divides the image into an $S \times S$ grid
- Each grid cell predicts B bounding boxes, Confidence for those boxes and C class probabilities.
- These predictions are encoded as an $S \times S \times (B \times 5 + C)$ tensor



How the YOLO algorithm works

- **Residual blocks**

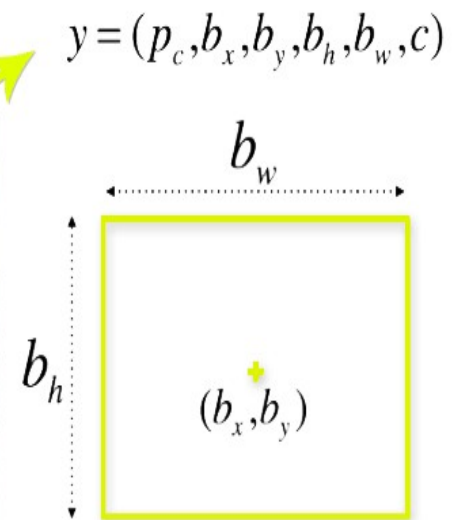
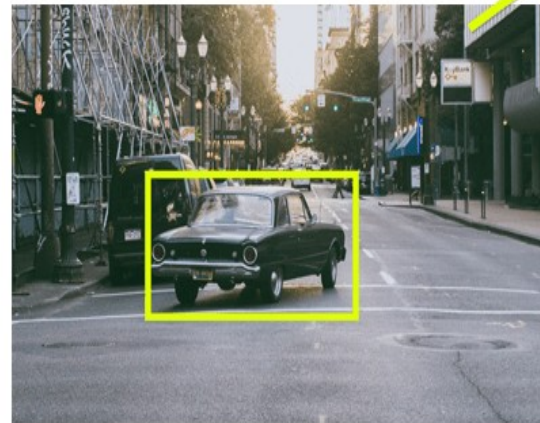
The input image is divided into grids, each grid has dimension $s \times s$. If an object center appears within a certain grid cell, then this cell will be responsible for detecting it.

- **Bounding box regression**

A bounding box is an outline that highlights an object in an image.

Used to predict width (b_w), height (b_h).

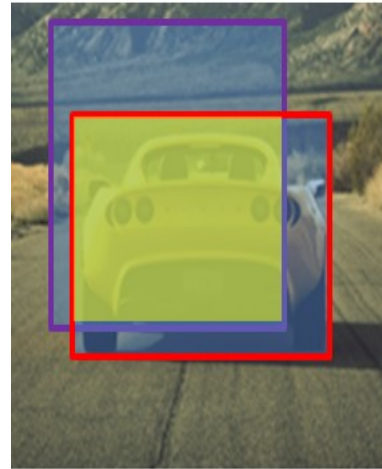
class (c), bounding box centre (b_x, b_y).



How the YOLO algorithm works

- **Intersection over union (IOU)**

if $IOU = 1$, then the predicted bounding box is same as real box. so that it eliminates all remaining boxes which are not equal to real boxes.



Intersection over union (IoU)

$$= \frac{\text{size of } \text{yellow box}}{\text{size of } \text{blue box}}$$

- **non-max suppression(NMS)**

discard all boxes with $P_c \leq 0.6$

for remaining boxes

pick largest P_c box and discard boxes with $IOU \geq 0.5$ with the box o/p in the previous step.



Applying
NMS



“Correct” if $IOU \geq 0.5$

Limitations & Advantages of YOLO

Limitations

- Struggles with small objects that appear in groups, such as flock of birds.
- Struggles to generalize with different aspects and ratio of objects
- Loss function treats errors as same for both small and large bounding boxes.

Advantages

- Speed
- High Accuracy
- Learning capabilities
- Robust object detection

Results

