



In511 Commandes Intelligentes

The second part of the course, October 2025

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The objective of
the second part
of the course

How to do object detection by training your own datasets and show the object on the camera by using vocal commands.



An ongoing project: Robot Voya

- Autonomous Movement
 - Ability to navigate environments using sensors (LIDAR, cameras).
 - Path planning and obstacle avoidance
- Perception System
 - AI-based vision for object classification
- Manipulation
 - Pick and place objects
- Human interaction
 - Human-tracking system with vision or sensors
 - Gesture recognition to interpret simple commands (e.g., waving to stop or follow)

Data Collection & Preparation

104 object to use in your experiments



Voice-Activated Real-Time Object Detection Using Custom Datasets

Model

Output

Input



“Show me Laptop”



Dataset Preparation



Model Training



Voice
model



highlighted only the
requested object

Data Collection & Preparation

- Create the data folder



In511/custom_dataset

Name	Type
 data	Fichier source Yaml
 images	File folder
 labels	File folder

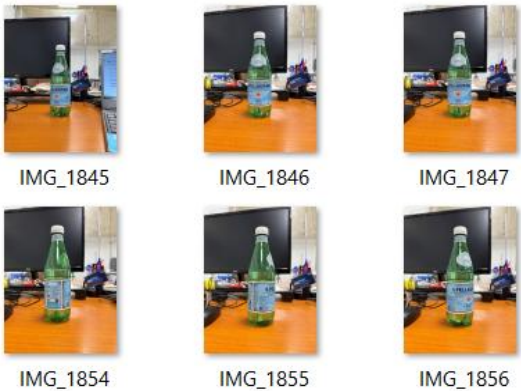
```
# data.yaml
path: ./custom_office_dataset # dataset root dir
train: images/train # train images (relative to path)
val: images/val # val images (relative to path)

# Classes
nc: 3 # number of classes
names: ['Bottle', 'Bee', 'Stapler']
```

In511/custom_dataset/images

Name	Date modified
 train	7/18/2025 11:31 AM
 val	7/18/2025 12:02 PM

In511/custom_dataset/images/train & In511/custom_dataset/images/val






your custom images



Data Collection & Preparation

- Create the data folder







In511/custom_dataset

Name	Type
 data	Fichier source Yaml
 images	File folder
 labels	File folder

In511/custom_dataset/label

Name	Date modified
 train	7/18/2025 11:31 AM
 val	7/18/2025 12:02 PM

In511/custom_dataset/labels/train

Name	Type
 classes	Document texte
 IMG_1845	Document texte
 IMG_1846	Document texte
 IMG_1847	Document texte
 IMG_1848	Document texte
 IMG_1849	Document texte

classes.txt




Fichier	Modifier	Affichage
<u>Bottle</u>		
<u>Bee</u>		
<u>Stapler</u>		

your custom object names



Data Collection & Preparation

- Create the data folder







In511/custom_dataset

Name	Type
 data	Fichier source Yaml
 images	File folder
 labels	File folder

In511/custom_dataset/label

Name	Date modified
 train	7/18/2025 11:31 AM
 val	7/18/2025 12:02 PM

In511/custom_dataset/labels/train

Name	Type
 classes	Document texte
 IMG_1845	Document texte
 IMG_1846	Document texte
 IMG_1847	Document texte
 IMG_1848	Document texte
 IMG_1849	Document texte

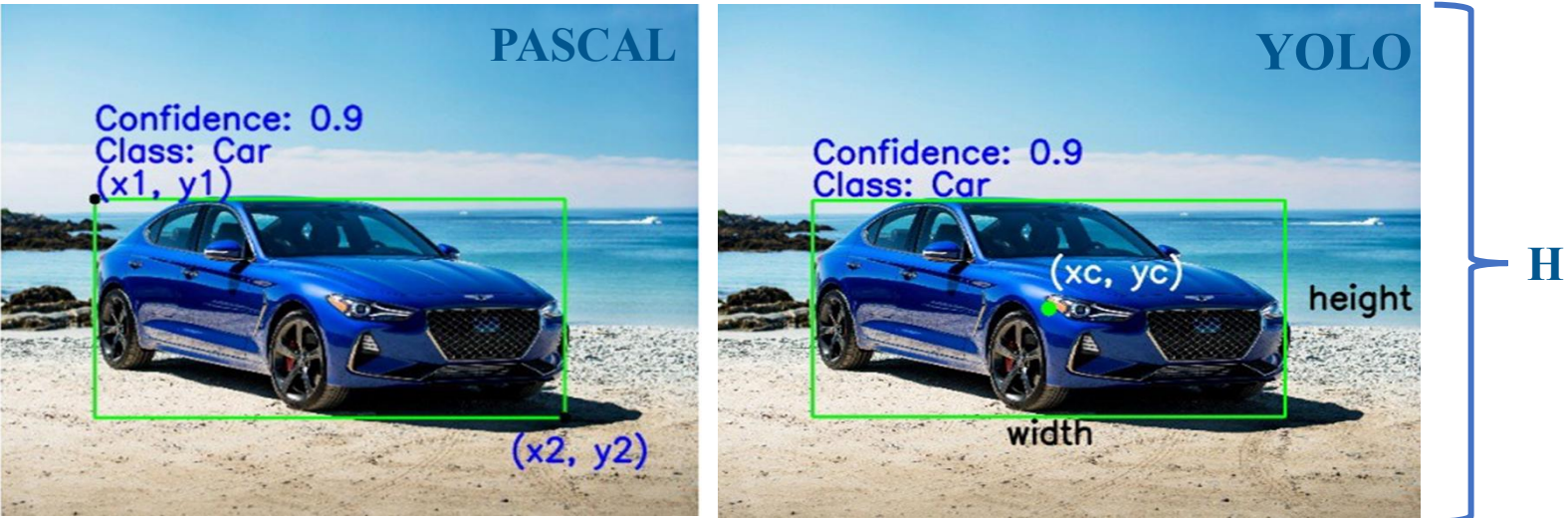
IMG_1845.txt

Fichier	Modifier	Affichage
0	0.502801 0.512430 0.188142 0.500000	

Class ID

Bounding Box
Values

What is Bounding Box? How to use it?



<https://nanonets.com/blog/image-processing-and-bounding-boxes-for-ocr/>

Format	Coordinates	Storage	Normalization	the center of the bounding box in pixels:	the normalized center coordinates:
PASCAL VOC	xmin, ymin, xmax, ymax	XML	No (pixels)	$X_{center} = \frac{X_1 + X_2}{2} \quad Y_{center} = \frac{Y_1 + Y_2}{2}$	$x_c = \frac{X_{center}}{W} \quad y_c = \frac{Y_{center}}{H}$
COCO	x, y, width, height	JSON	No (pixels)	W x H: the size of the Image, width: the width of the bounding box, height: the height of the bounding box.	
YOLO	x_center, y_center, width, height	TXT	Yes (0–1)	(X₁, Y₁): the X and Y coordinates of the top left corner of the rectangle. (X₂, Y₂): the X and Y coordinates of the bottom right corner of the rectangle. (X_c, Y_c): the normalized coordinates of the center of the bounding box.	

Data Collection & Preparation

Capture **100 images** of the custom object using your phone.

- **75 images** will be used for training.
- **25 images** will be used for validation.
- Make sure to capture the object from **different angles** and at **various distances** to improve dataset diversity.



Example images of a custom object

Data Collection & Preparation

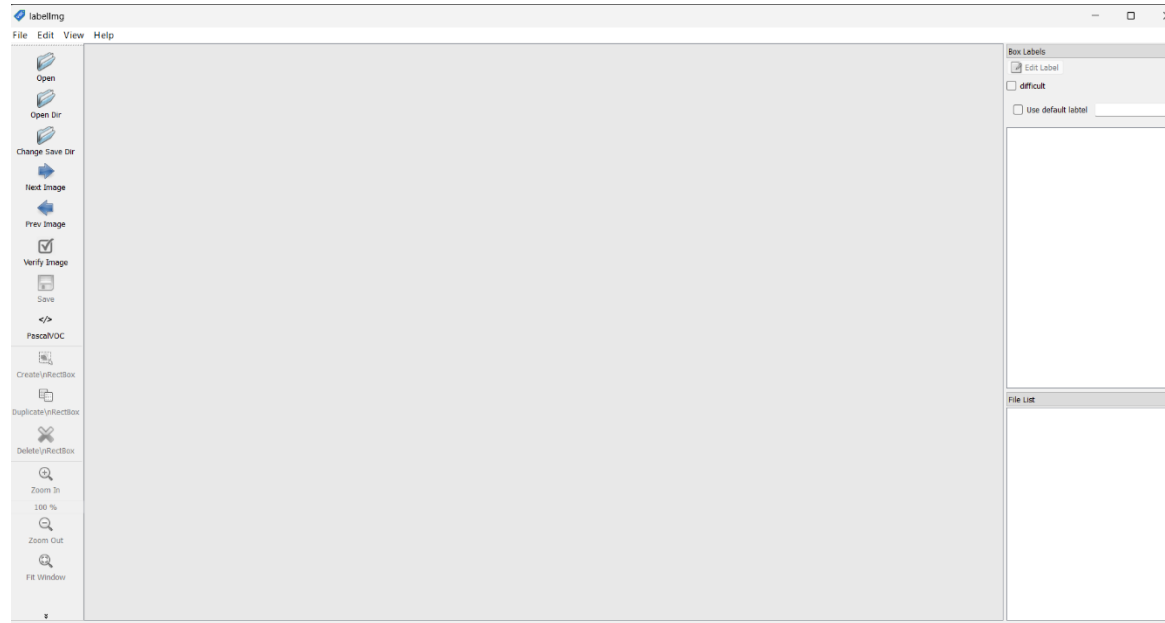
- If your images are in **HEIC format**, convert them to **PNG format** using the provided example code on Moodle: **heicToPngConverter.py** or you can find a heicToJpeg code online.

- Create labels for your dataset

You can download the desktop version of the labeling tool **LabelImg** from the following link:

<https://github.com/HumanSignal/labelImg/releases>

LabelImg Application

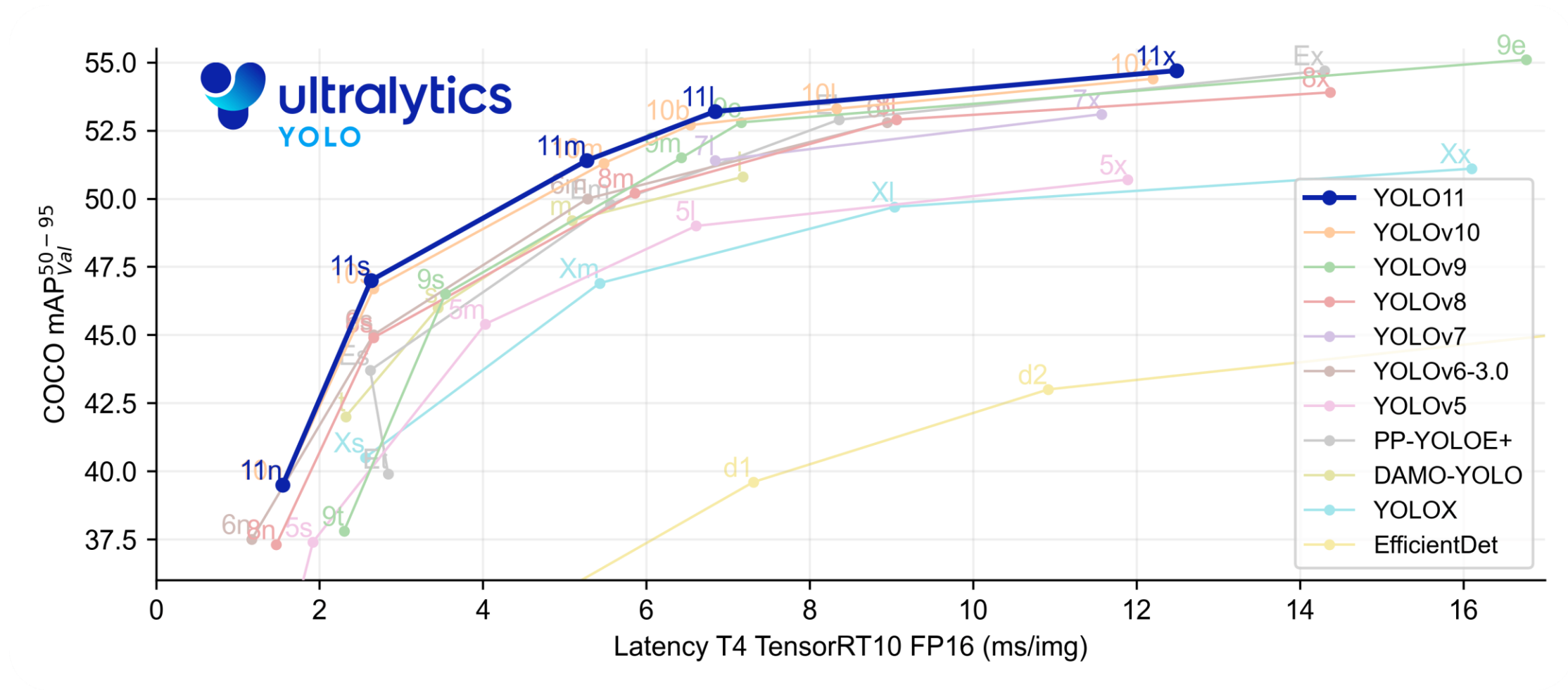


Model Training - History of YOLO

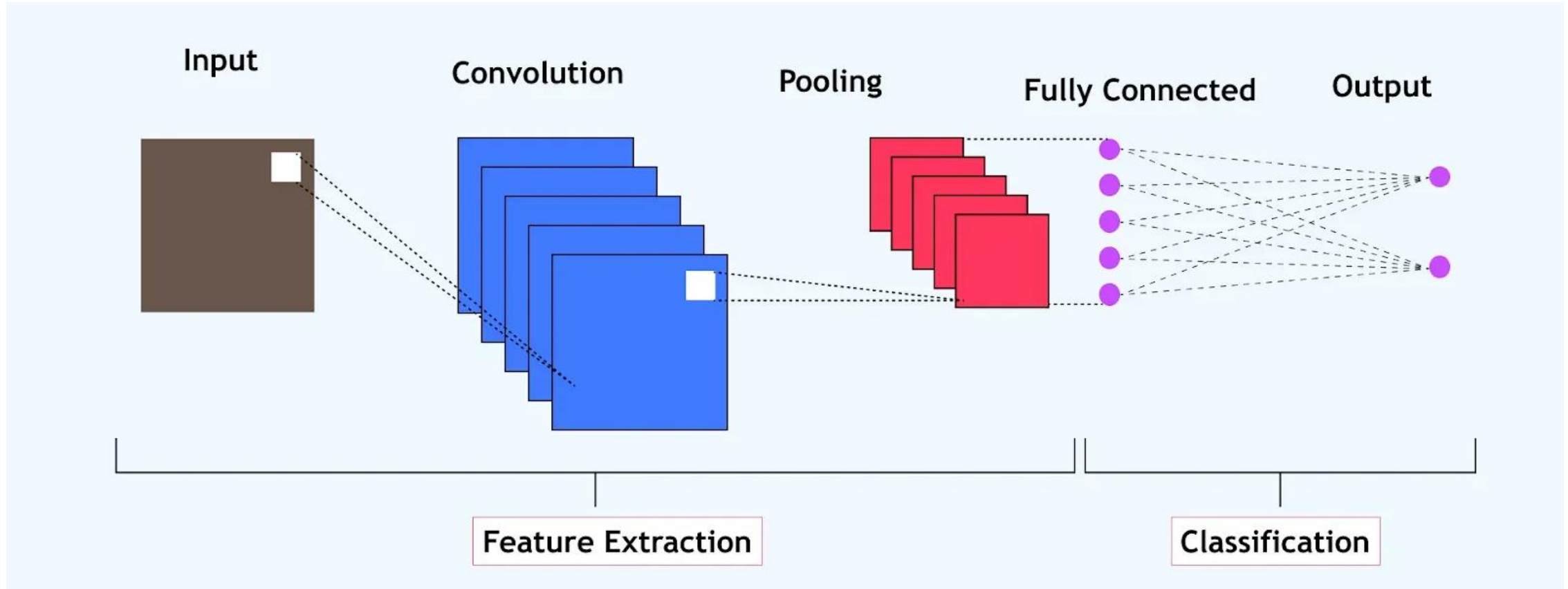
Year	Version	Highlights
2016	YOLOv1 – <i>Redmon et al.</i>	Introduced YOLO in “ You Only Look Once: Unified, Real-Time Object Detection ” (CVPR 2016), establishing the foundation of single-pass object detection.
2017	YOLOv2 / YOLO9000	Added anchor boxes, batch normalization, and real-time detection of 9,000 classes.
2018	YOLOv3	Adopted Darknet-53 backbone, multi-scale detection, and improvements in accuracy.
2020–2023	YOLOv4, v5, v6, v7, v8	Continuous improvements in speed and accuracy. e.g., YOLOv7 set new state-of-the-art (SOTA) results.
2024	YOLOv10	Introduced NMS-free training, dual-assignment strategy, and holistic efficiency improvements.
September 2024	YOLO11	Released by Ultralytics at YOLO Vision 2024 (YV24); features architectural refinements (C3k2, C2PSA blocks), fewer parameters, improved speed, and multi-task support.
February 18, 2025	YOLOv12 – <i>Tian et al.</i>	Attention-centric architecture (Area Attention, R-ELAN, FlashAttention). Achieves SOTA mAP with very low latency—e.g., YOLOv12-N hits 40.6% mAP in 1.64 ms , outperforming prior versions.

Reference for ULTRALYTICS: <https://docs.ultralytics.com/models/yolo11/>

Performance of YOLO Versions



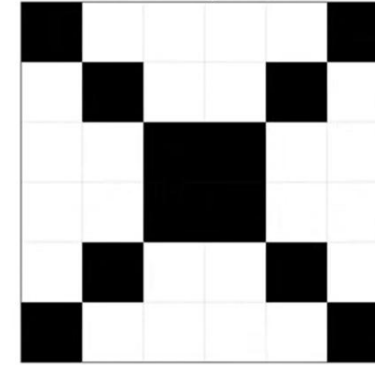
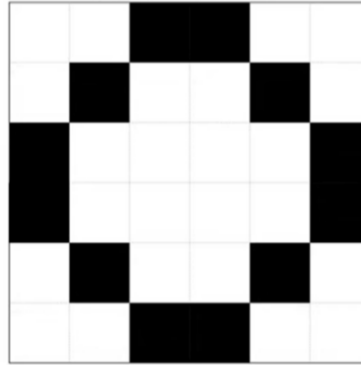
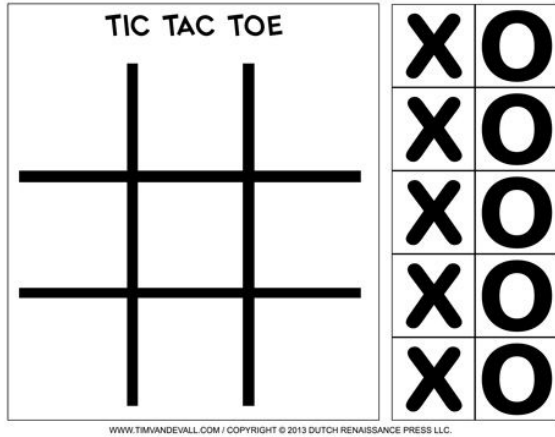
Basic Architecture of Convolutional Neural Networks (CNNs)



How does Convolutional Neural Networks (CNNs) work?

Example

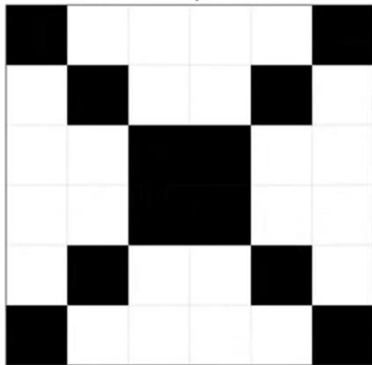
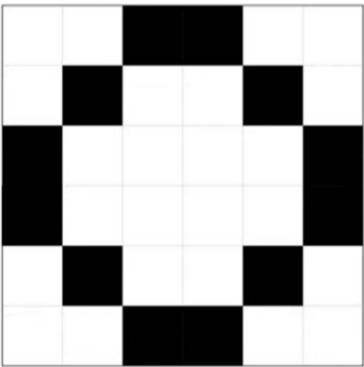
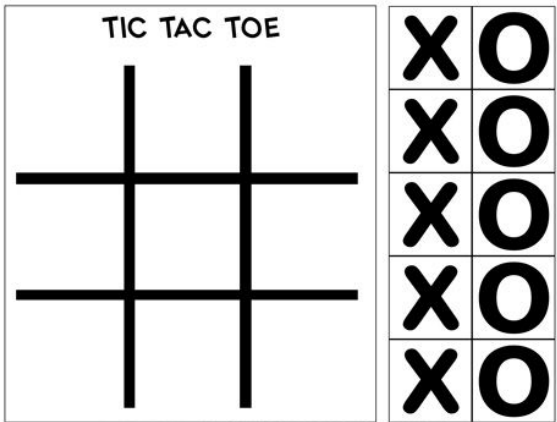
Pixel Representation



How does Convolutional Neural Networks (CNNs) work?

Example

Pixel Representation



0 for white, 1 for black pixels

0	0	1	1	0	0
0	1	0	0	1	0
1	0	0	0	0	1
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0

1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0
0	0	1	1	0	0
0	1	0	0	1	0
1	0	0	0	0	1

How does Convolutional Neural Networks (CNNs) work?

Example

0	0	1	1	0	0
0	1	0	0	1	0
1	0	0	0	0	1
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0



(3x3) Kernel Matrices (Filters)

0	0	1
0	1	0
1	0	0

1	0	1
0	0	1
1	1	0

0	0	1	1	0	0
0	1	0	0	1	0
1	0	0	0	0	1
1	0	0	0	0	1
0	1	0	0	1	0
0	0	1	1	0	0

Dot Product



Filter

0	0	1
0	1	0
1	0	0

$$\begin{aligned} & (0 \times 0) + (0 \times 0) + (1 \times 1) \\ & + (0 \times 0) + (1 \times 1) + (0 \times 0) \\ & + (1 \times 1) + (0 \times 0) + (0 \times 0) \end{aligned}$$

= 3

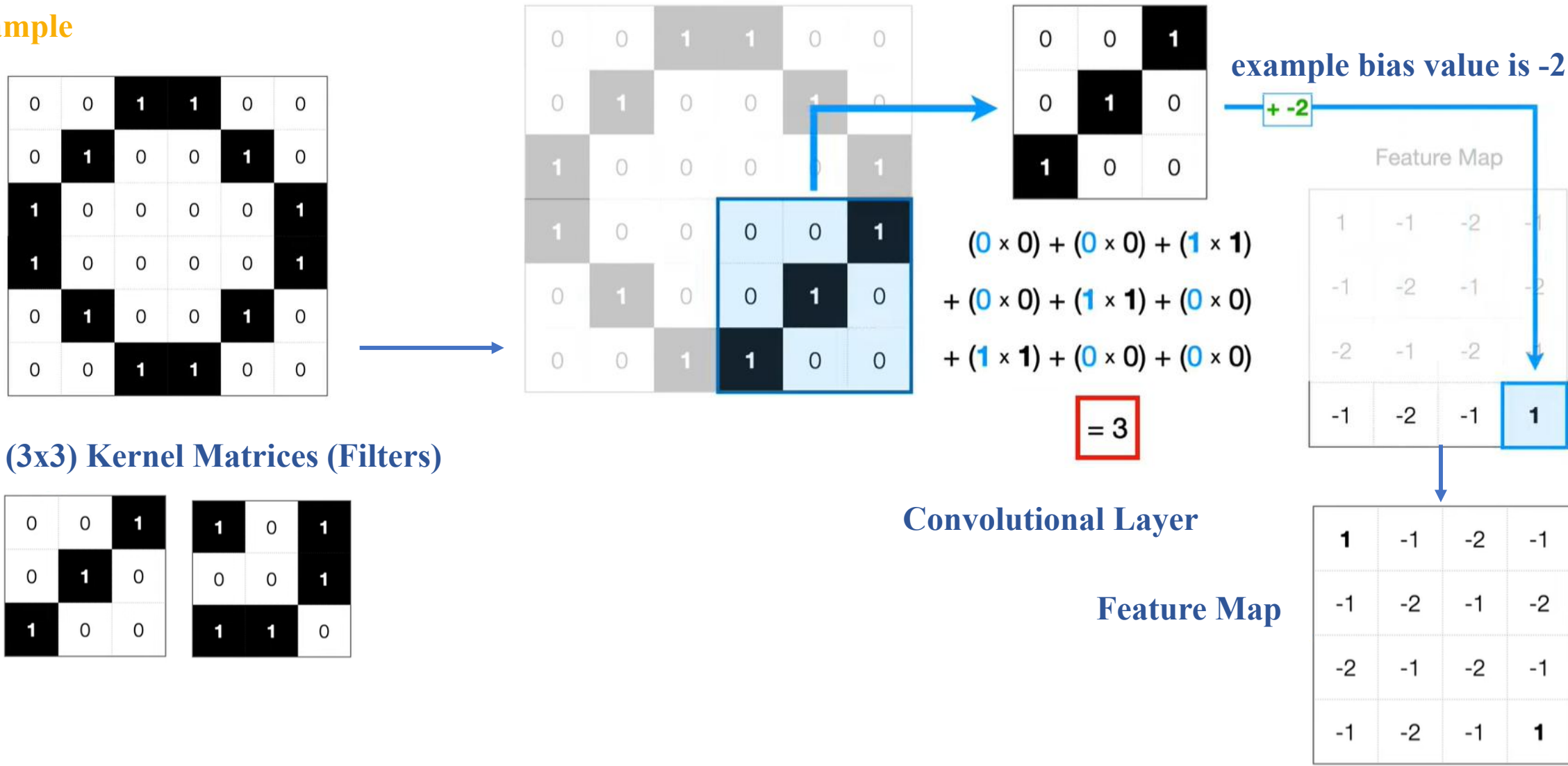
Convolutional Layer

In Python, for example;

```
nn.Conv2d(in_channels=1, out_channels=32, kernel_size=3)
```

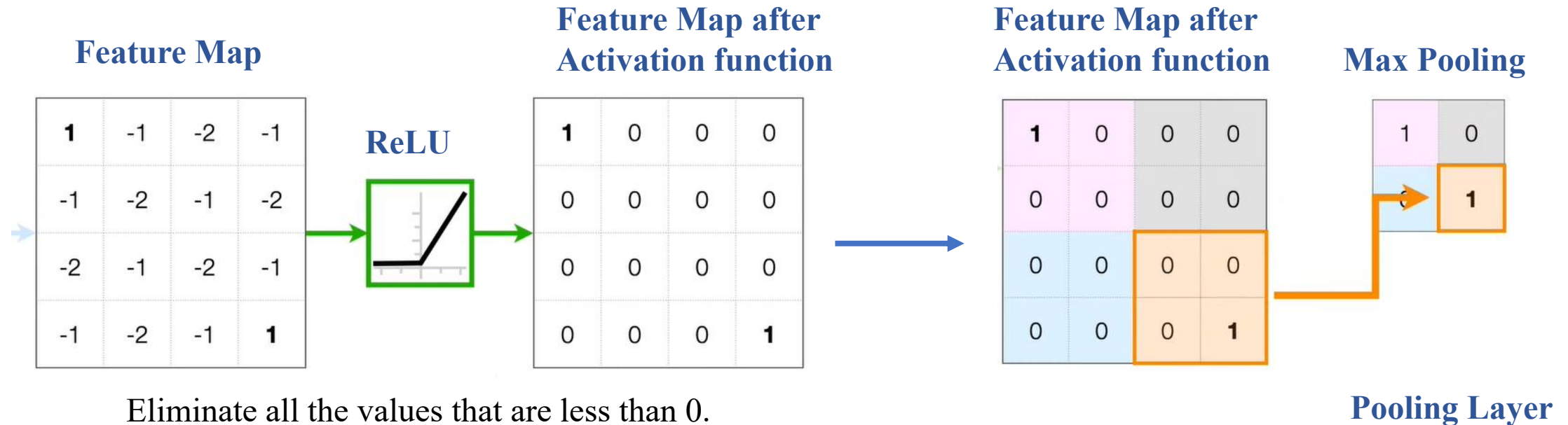
How does Convolutional Neural Networks (CNNs) work?

Example



How does Convolutional Neural Networks (CNNs) work?

Example



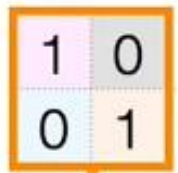
ReLU - Rectified linear unit

<https://www.youtube.com/watch?v=HGwBXDKFk9I>

How does Convolutional Neural Networks (CNNs) work?

Example

Max Pooling
result

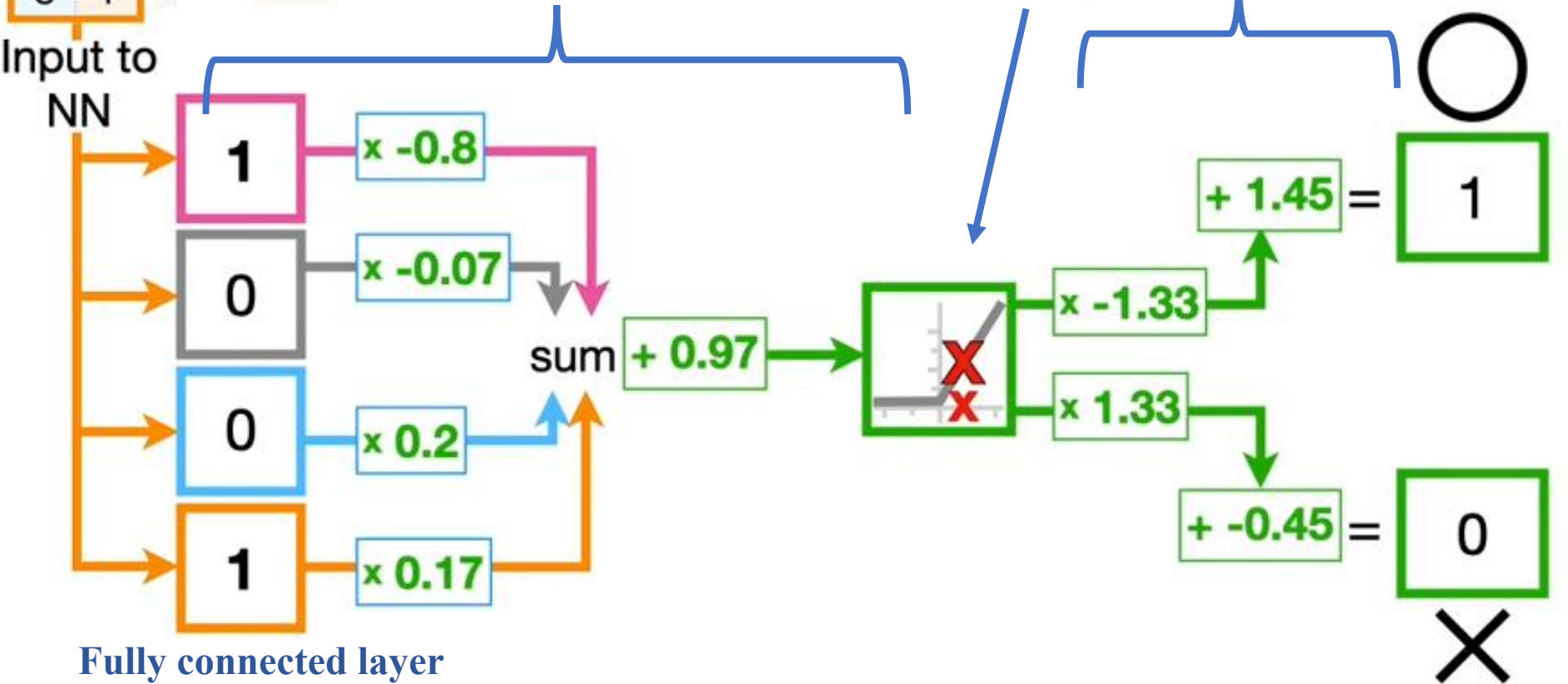


Input to
NN

$(1 \times -0.8) + (0 \times -0.07) + (0 \times 0.2) + (1 \times 0.17) + 0.97 = 0.34$

$(0.34 \times -1.33) + 1.45 = 1$

Fully connected layer



How does Region-based CNN (R-CNN) work?

Example Image



How to handle the same object with different sizes?

Crops of different size



Crops of downscale version of the image



<https://arxiv.org/pdf/1311.2524>

<https://www.youtube.com/watch?v=nJzQDpppFj0&t=1s>

<https://www.youtube.com/watch?v=5DvljLV4S1E>

How does Region-based CNN (R-CNN) work?

Example Image



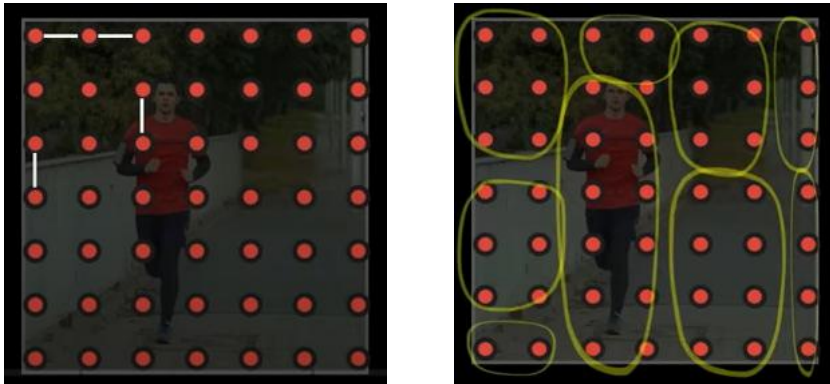
How to choose regions to pass through the CNN?

Selective Search

Step 1: segment image into regions

Step 2: merge similar regions to create larger regions

Step 1
check distance between pixels



How does Region-based CNN (R-CNN) work?

Example Image



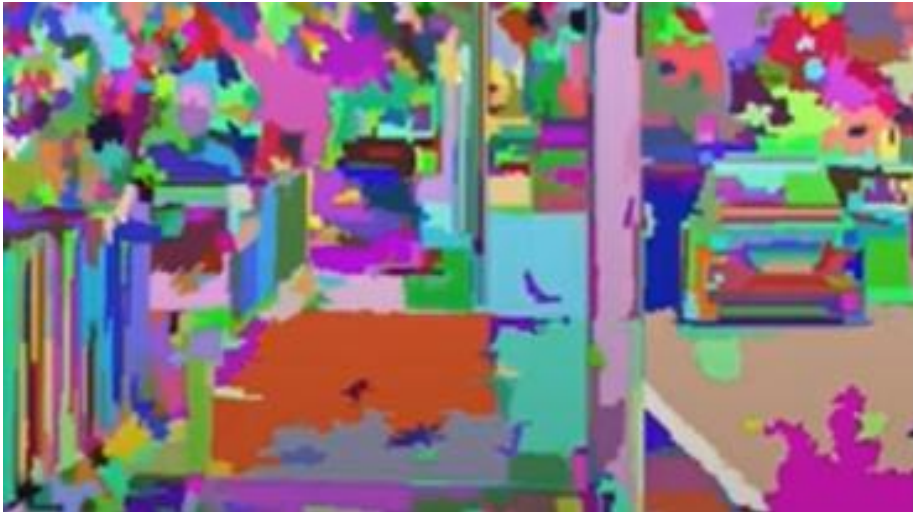
How to choose regions to pass through the CNN?

Selective Search

Step 1: segment image into regions

Step 2: merge similar regions to create larger regions

Step 2
check similarity based on color, texture, size, shape



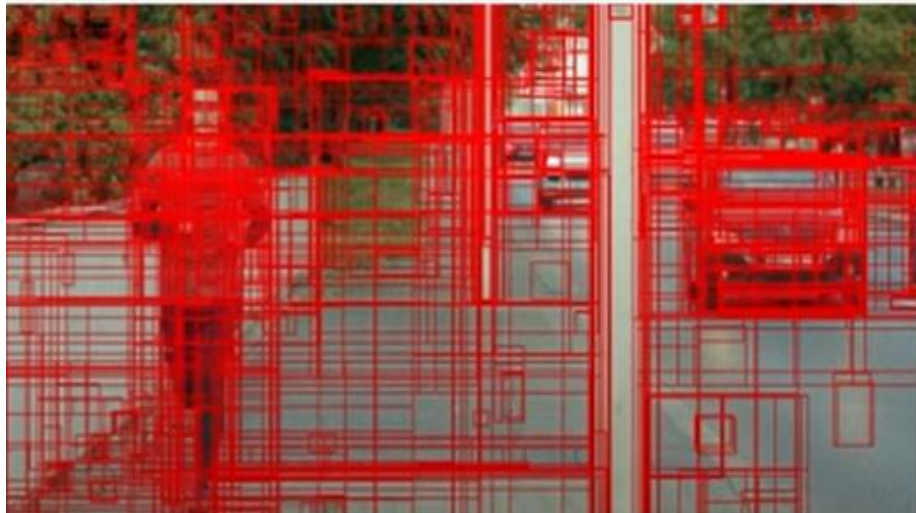
How does Region-based CNN (R-CNN) work?

Example Image

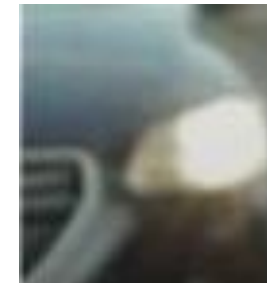


Any arbitrary shape proposed regions, how to pass them through the CNN?

Fix size region should be given as an input
Label for every chosen regions



1600 region proposals



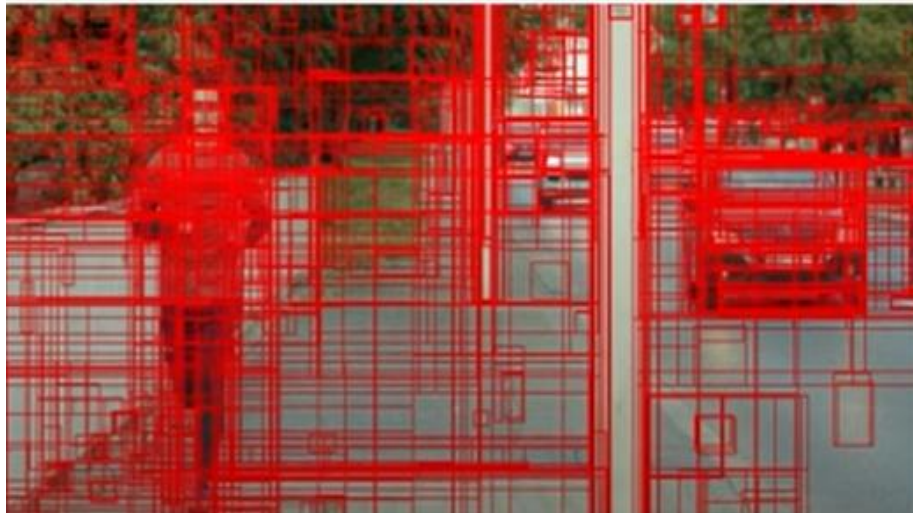
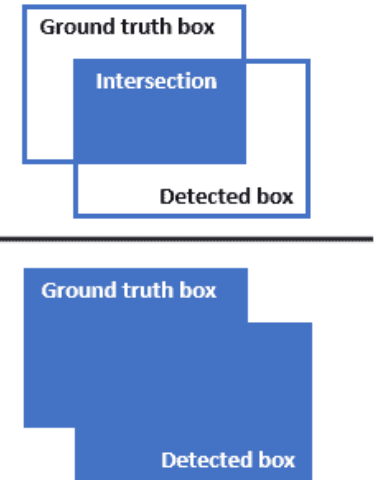
How does Region-based CNN (R-CNN) work?

Example Image

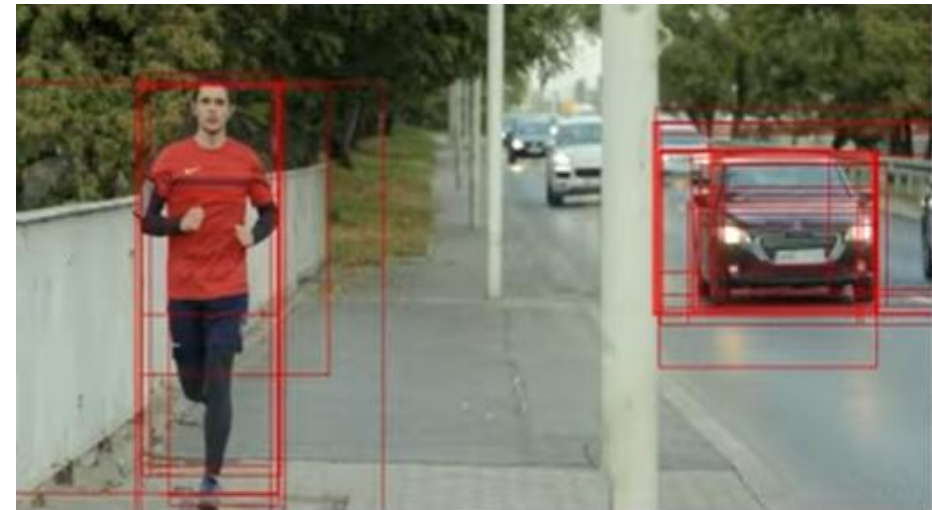


Intersection over Union

$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}} =$$



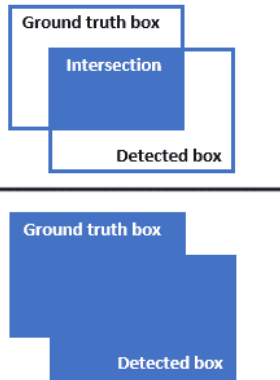
1600 region proposals



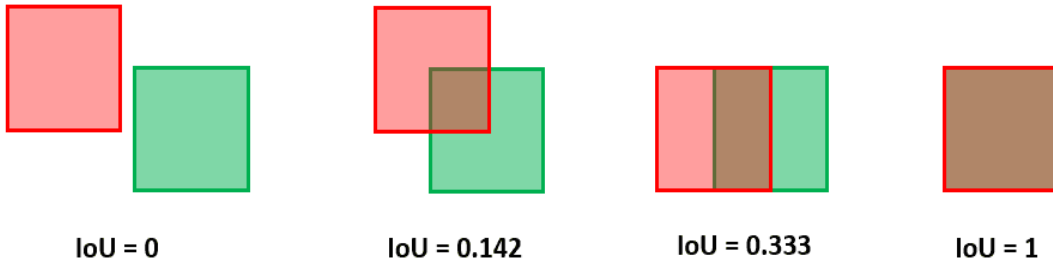
Solution: Non-maximum suppression

Intersection over Union (IoU)

IoU measures the accuracy of the detections

$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}} = \frac{\text{Intersection}}{\text{Union}}$$


Examples of different IoU values



<https://www.baeldung.com/cs/object-detection-intersection-vs-union>

The IoU-based loss functions

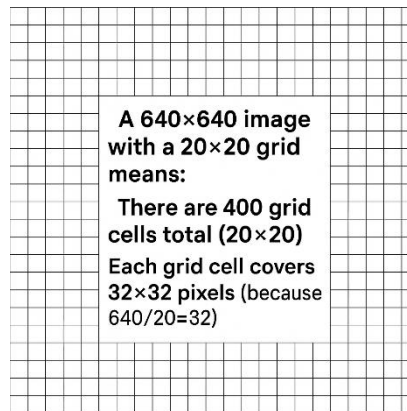
GIoU (Generalized IoU)

DIoU (Distance IoU)

CIoU (Complete IoU)

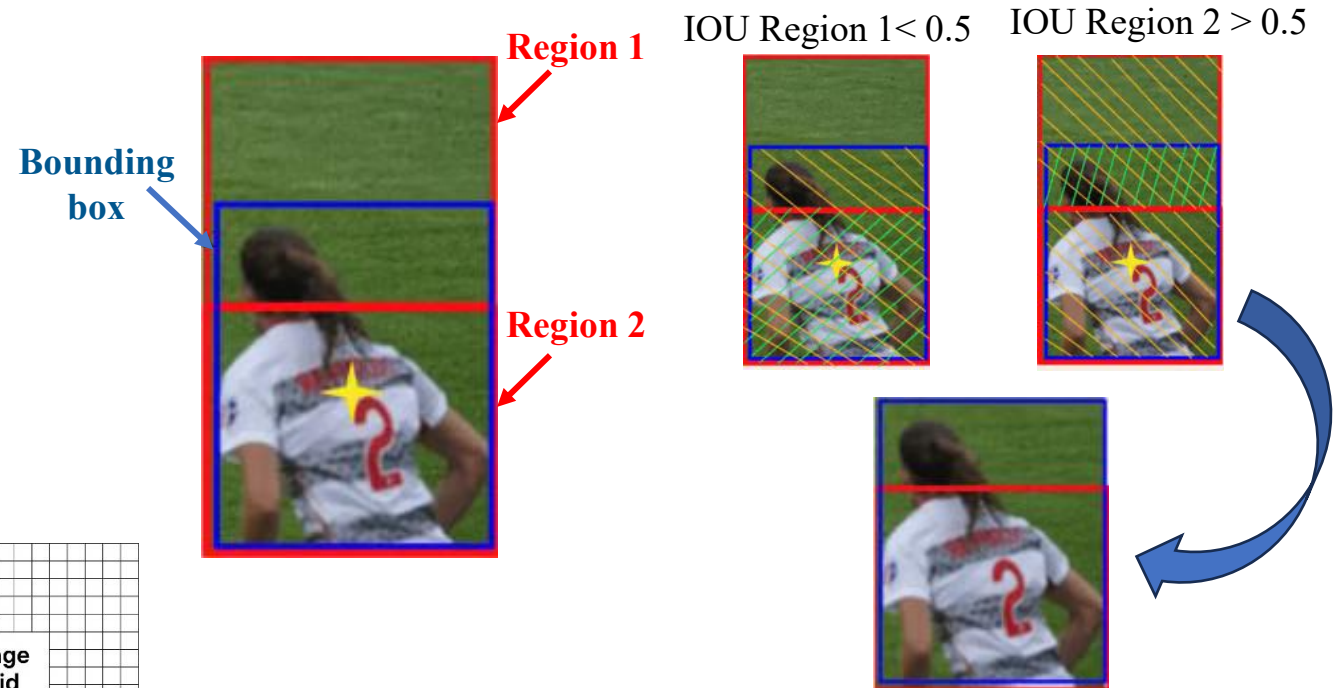
SIoU (Scalable IoU)

} YOLOv8



Example

- The user can define an IOU selection threshold, for instance, 0.5.
- YOLO computes the IOU of each grid cell with the formula on the left side.
- It considers only IOU > threshold.

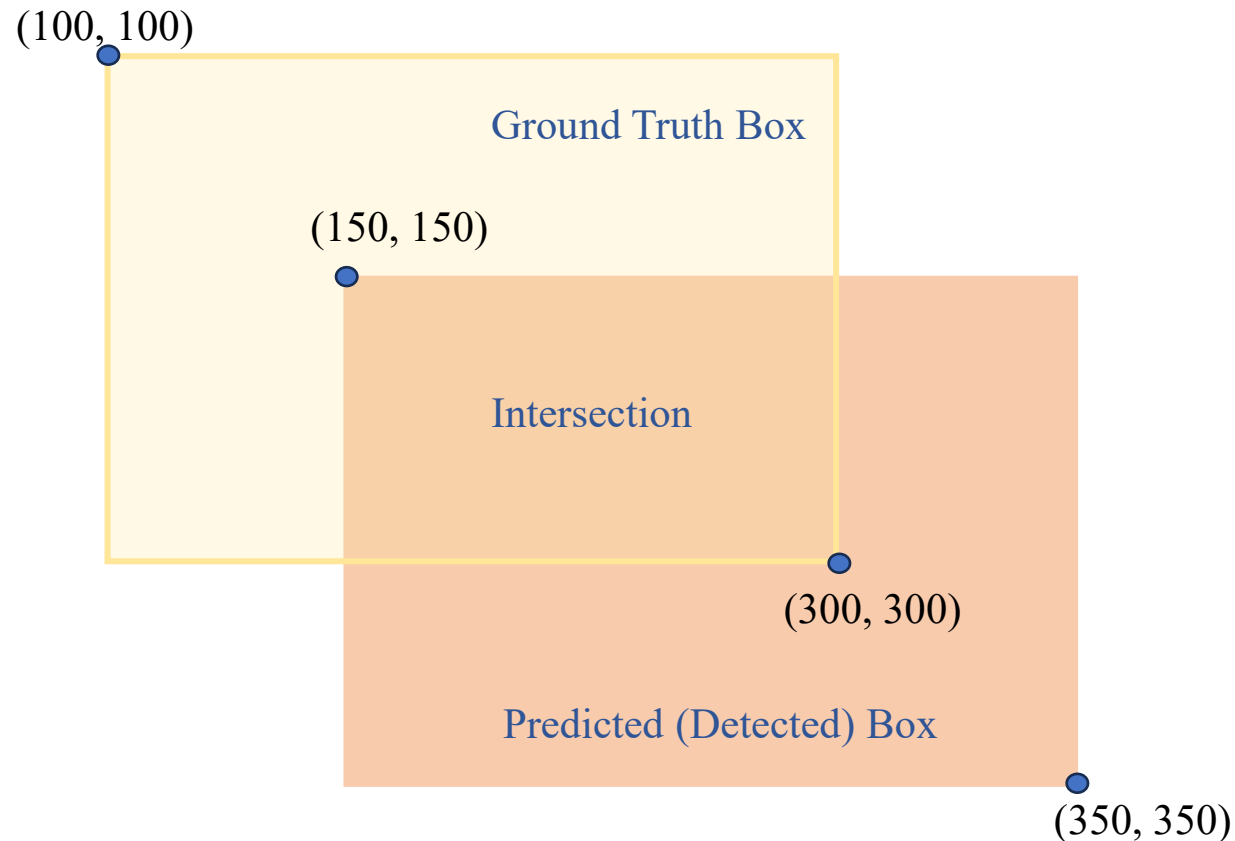


<https://www.datacamp.com/blog/yolo-object-detection-explained>

Intersection over Union (IoU)

IoU measures the accuracy of the detections

$$\text{IoU} = \frac{\text{Area of Overlap}}{\text{Area of Union}}$$



$$\text{Area of Ground Truth Box} = 200 \times 200 = 40,000 \text{ pixels}^2$$

$$\text{Area of Predicted Box} = 200 \times 200 = 40,000 \text{ pixels}^2$$

$$\text{Area of Intersection} = 150 \times 150 = 22,500 \text{ pixels}^2$$

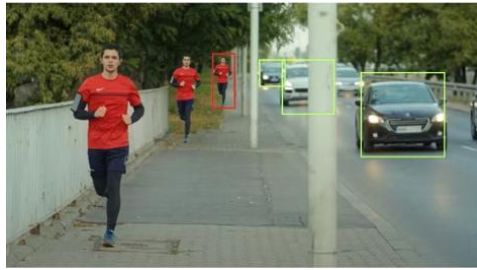
$$\text{Union} = \text{Area}(\text{GT}) + \text{Area}(\text{P}) - \text{Intersection}$$

$$\text{Union} = 40,000 + 40,000 - 22,500 = 57,500$$

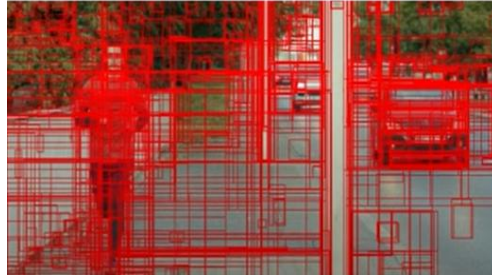
$$\text{IoU} = \text{Intersection over Union} = \frac{22,500}{57,500} \approx 0.391$$

How does Region-based CNN (R-CNN) work?

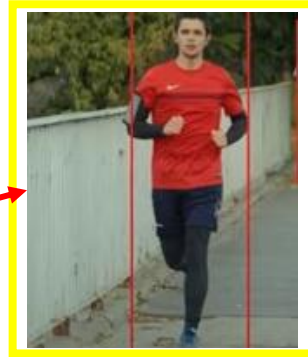
Example Image



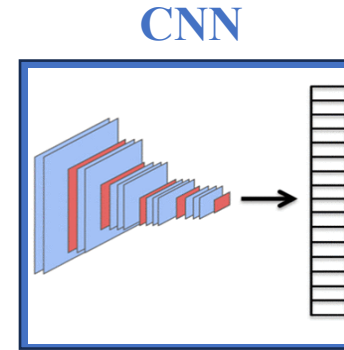
Input



Region Proposals



Resized
Region
Image



Feature Vector

Classification	Result
SVM → Dog	Yes
SVM → Tree	No
SVM → Car	No
SVM → Bike	No

Train with $\text{IoU} \geq 0.5$
with ground truth
boxes of that class

SVM – Support vector machine – classification method

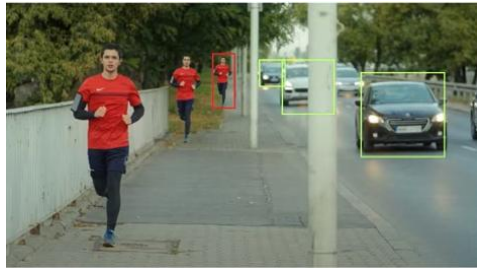
Girshick et. al (2014) “Rich feature hierarchies for accurate object detection and semantic segmentation”

<https://arxiv.org/pdf/1311.2524>

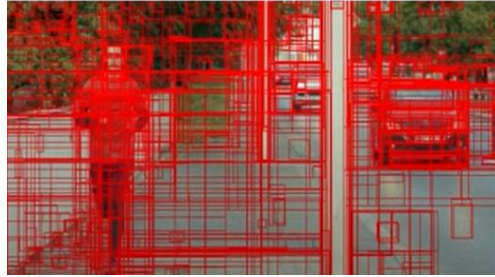
Source: Redmon and <https://www.youtube.com/watch?v=svn9-xV7wjk&t=170s>

How does Region-based CNN (R-CNN) work?

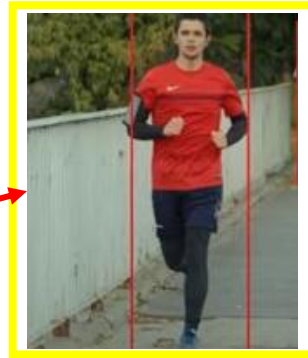
Example Image



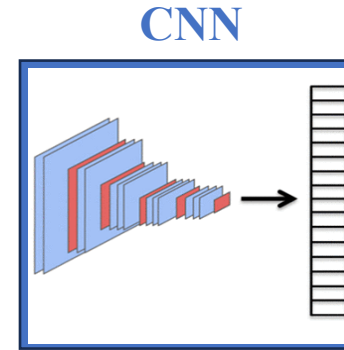
Input



Region Proposals

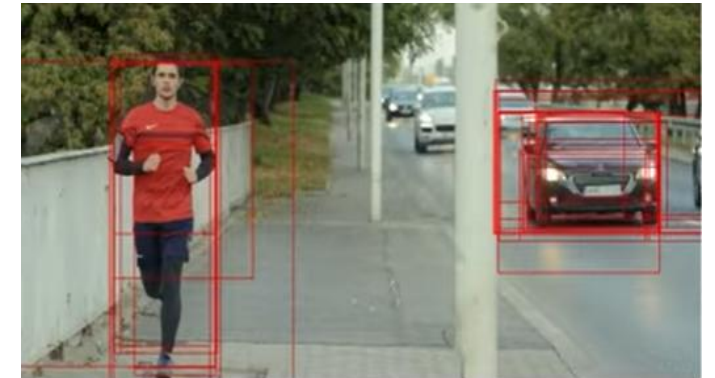


Resized
Region
Image



Feature Vector

Classification	Result
SVM → Dog	Yes
SVM → Tree	No
SVM → Car	No
SVM → Bike	No



How to evaluate the model R-CNN?

- apply **Non-Maximum Suppression (NMS)** to remove duplicate detections. → If $IoU(P_1, P_2) > Threshold$:
- evaluate predictions using **Average Precision (AP)** per class and **Mean Average Precision (mAP)** across classes.

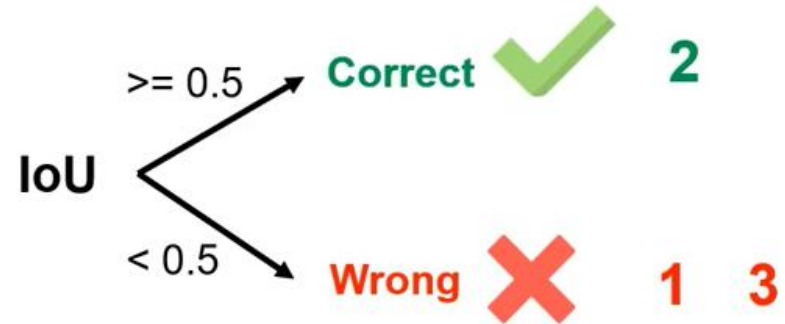
$$P = \operatorname{argmax}(C(P_1), C(P_2))$$

How does Region-based CNN (R-CNN) work?

Example Image

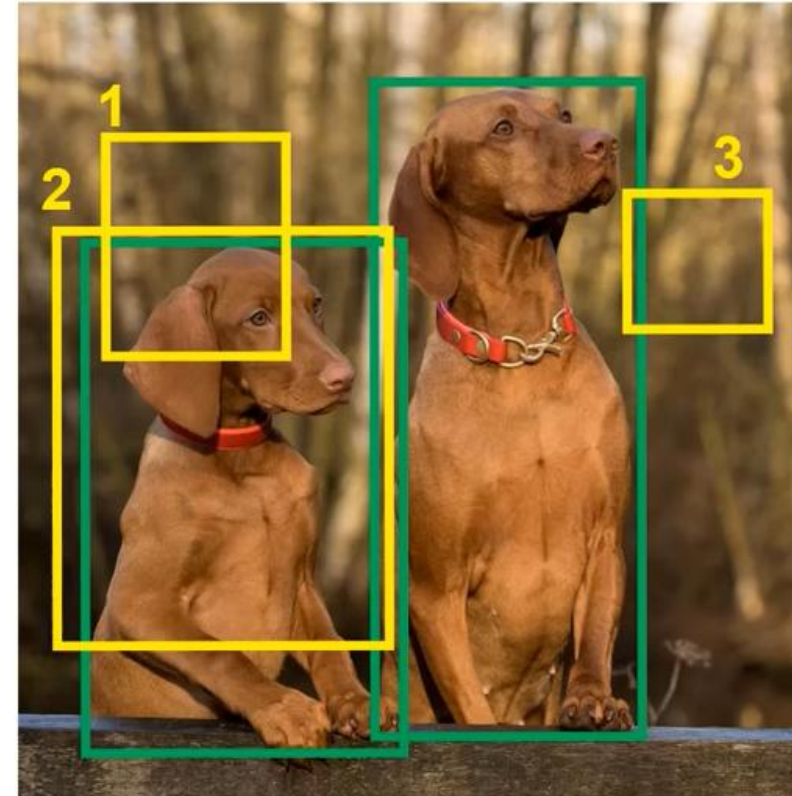
Mean Average Precision (mAP)

Which predicted bounding boxes are correct?



Precision $\frac{1}{3}$

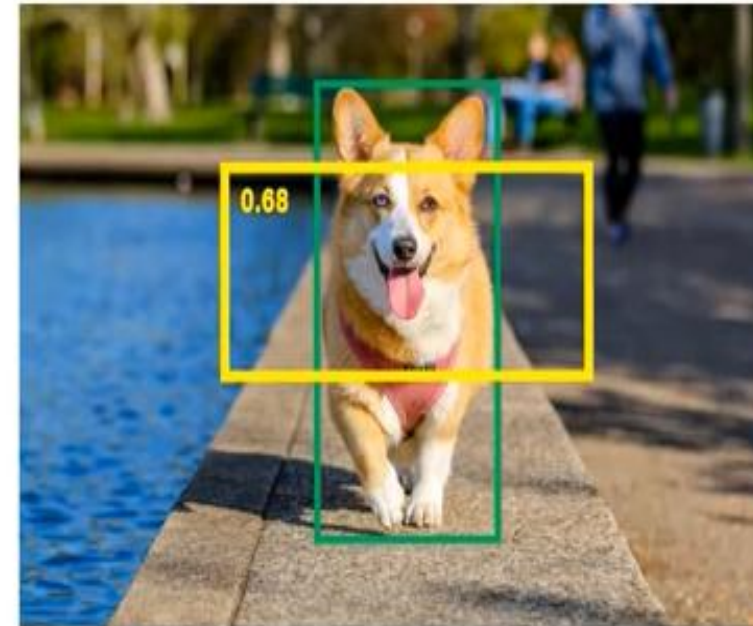
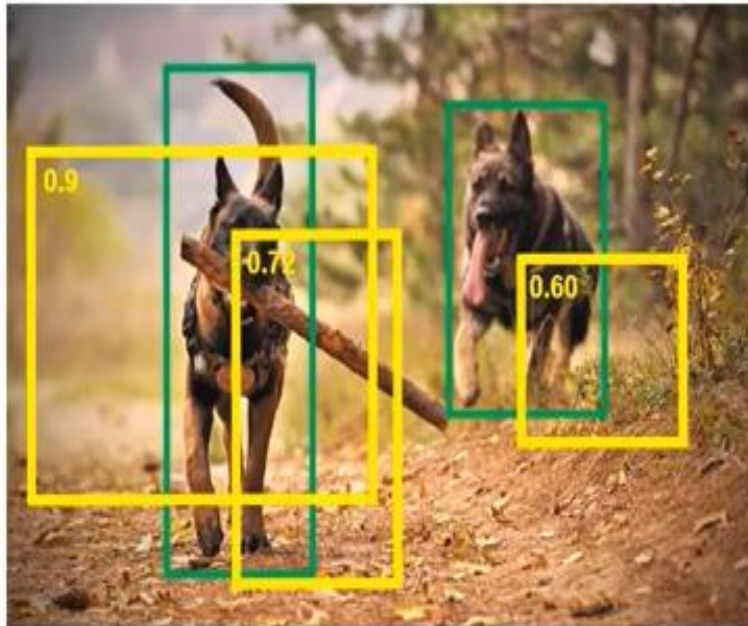
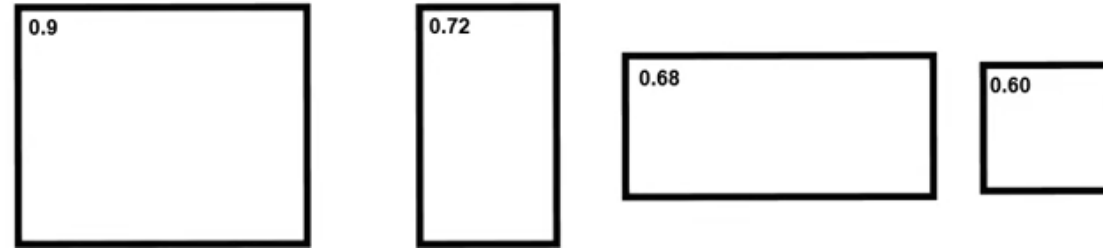
Recall $\frac{1}{2}$



How does Region-based CNN (R-CNN) work?

Example Image

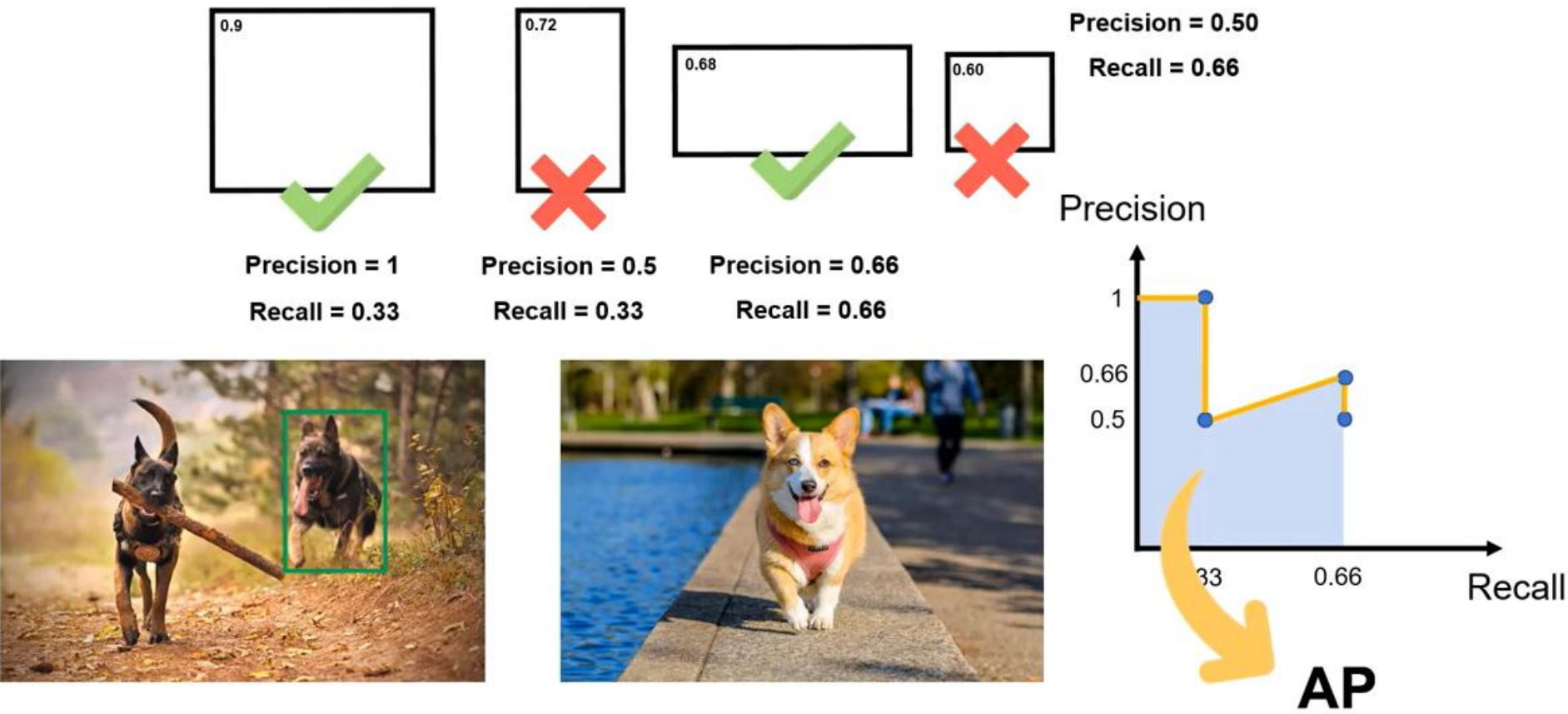
Mean Average Precision (mAP)



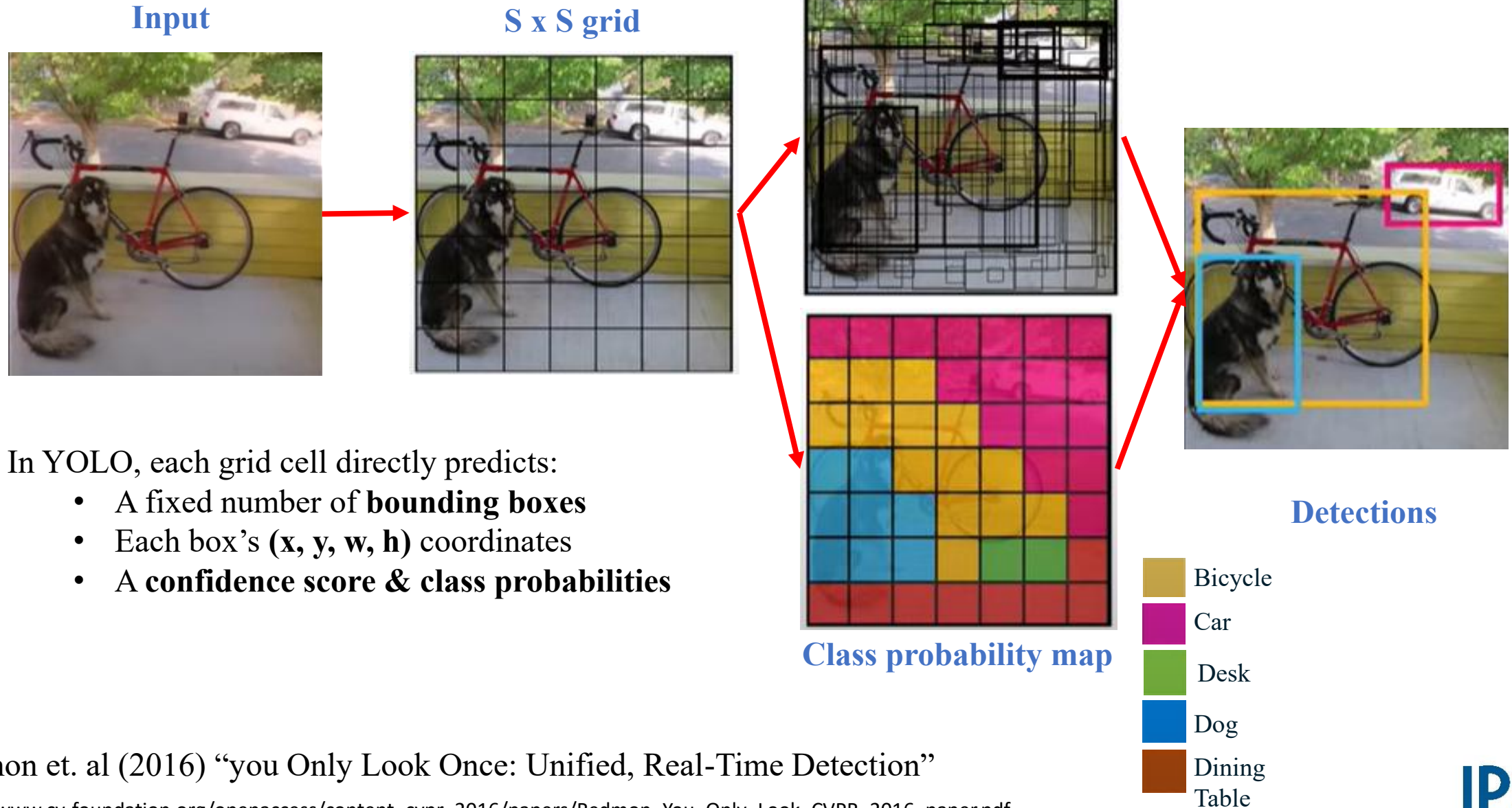
How does Region-based CNN (R-CNN) work?

Example Image

Mean Average Precision (mAP)



How does YOLO Work?



Redmon et. al (2016) “you Only Look Once: Unified, Real-Time Detection”

https://www.cv-foundation.org/openaccess/content_cvpr_2016/papers/Redmon_You_Only_Look_CVPR_2016_paper.pdf

Source: <https://www.youtube.com/watch?v=svn9-xV7wjk&t=170s>

How does YOLO Work?



S x S grid

Output Vector Length

$$B \times 7 + n$$

Class probabilities

$$[p(c_1), p(c_2), \dots, p(c_n)]$$

Bounding box predictions

$$[x_1, y_1, \sqrt{w_1}, \sqrt{h_1}, C_1]$$

$$[x_2, y_2, \sqrt{w_2}, \sqrt{h_2}, C_2]$$

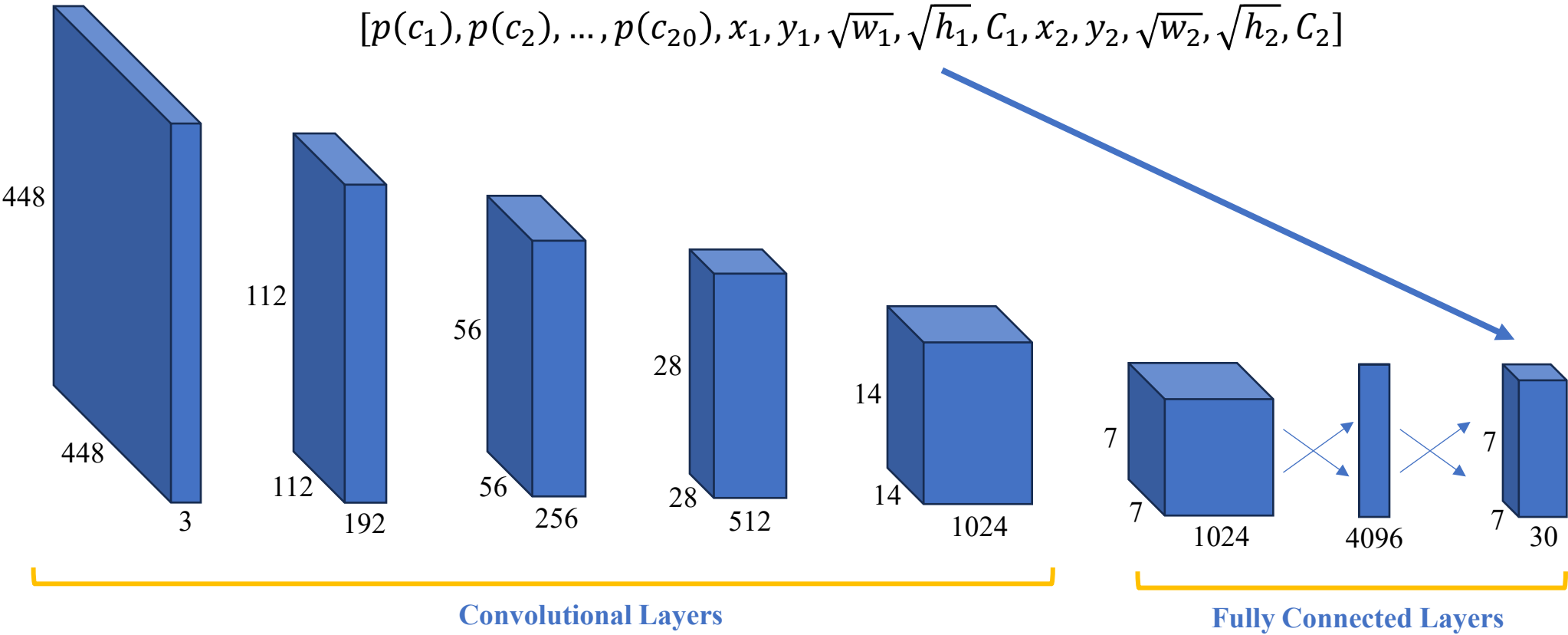
...

$$[x_B, y_B, \sqrt{w_B}, \sqrt{h_B}, C_B]$$

n is the number of object classes

B is the number of bounding boxes

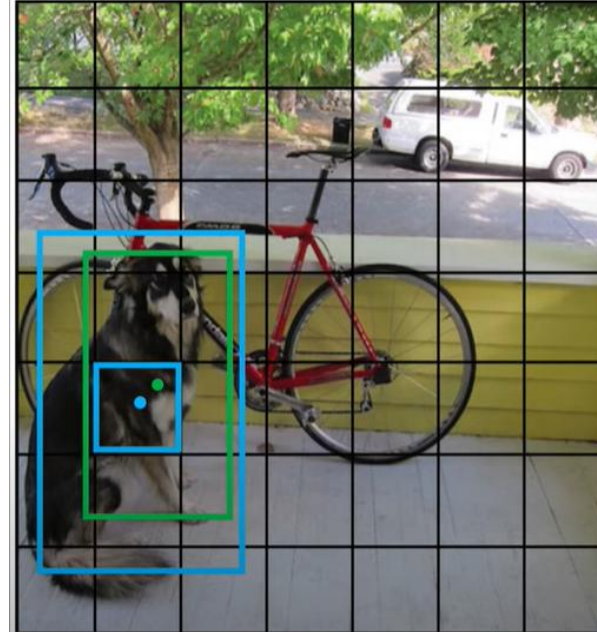
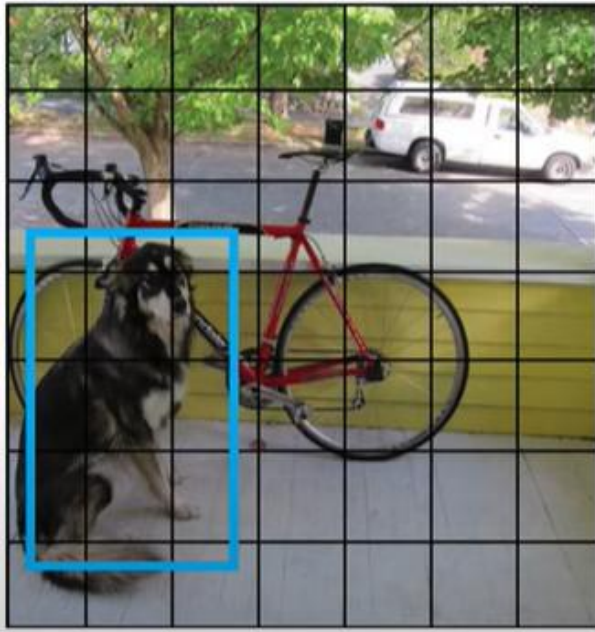
How does YOLO Work?



Source: <https://www.youtube.com/watch?v=svn9-xV7wjk&t=170s>

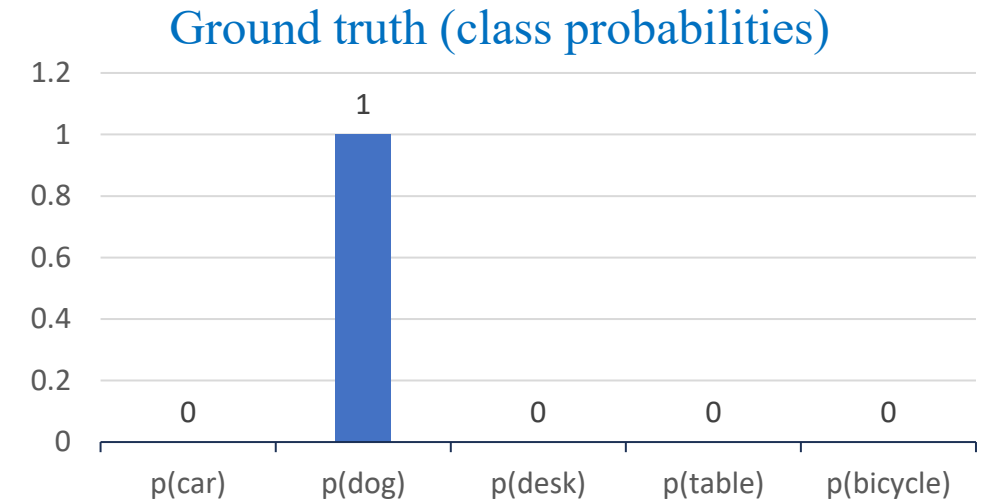
How does YOLO Work?

Ground truth (class probabilities)



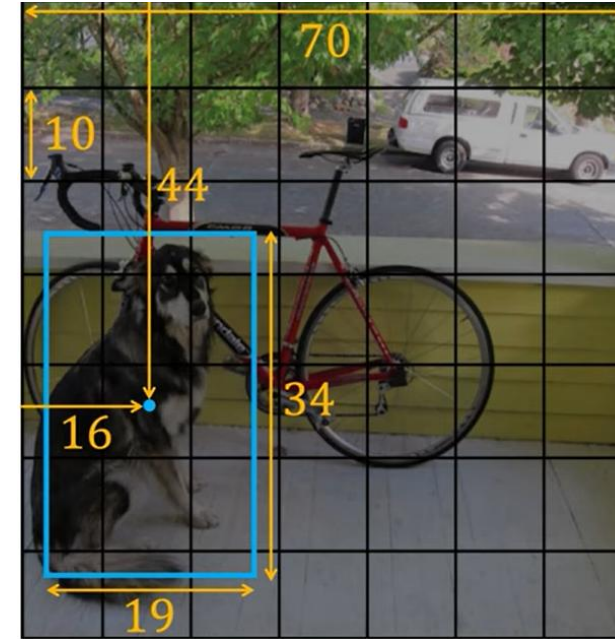
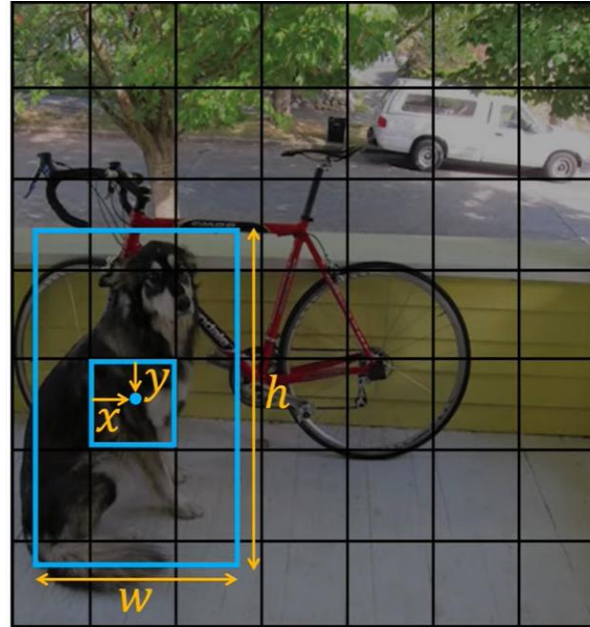
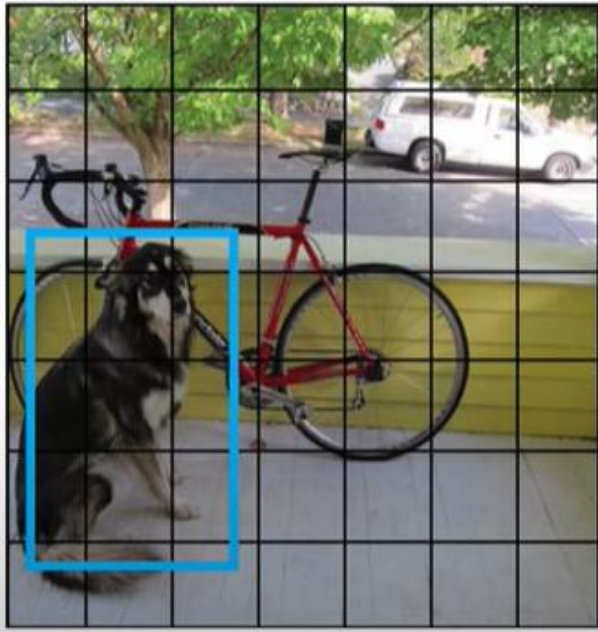
Let's predict this bounding box

Green box is our prediction



How does YOLO Work?

Ground truth (box coordinates)



$$x = \frac{16 \div 10}{10} = 0.60$$

$$y = \frac{44 \div 10}{10} = 0.40$$

$$w = \frac{19}{70} = 0.27$$

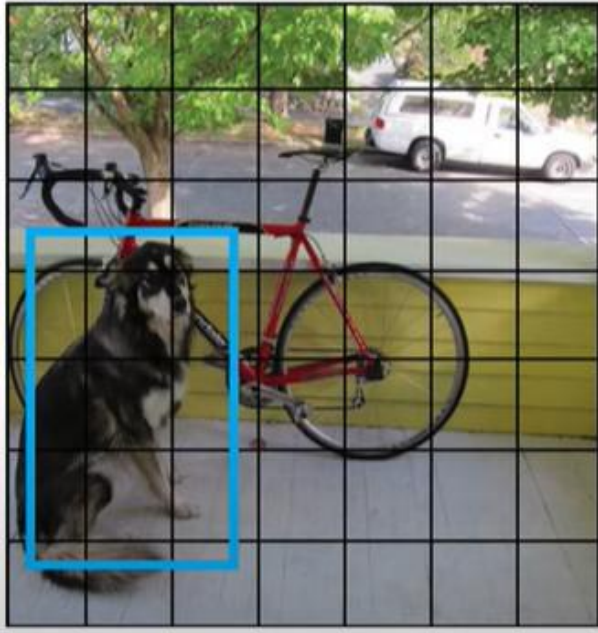
$$h = \frac{34}{70} = 0.49$$

Let's predict this bounding box

Calculate the ground truth and predicted boxes
by using the coordinates


How does YOLO Work?


Ground truth (confidence)

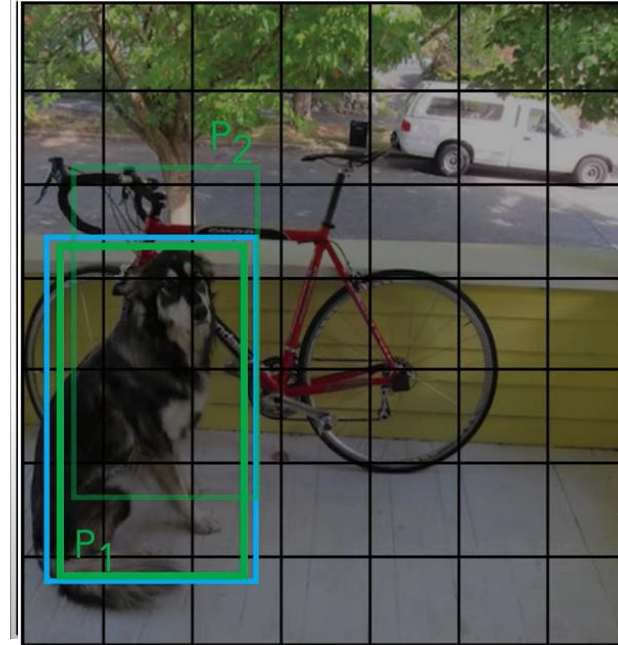


Let's predict this bounding box



 = 0

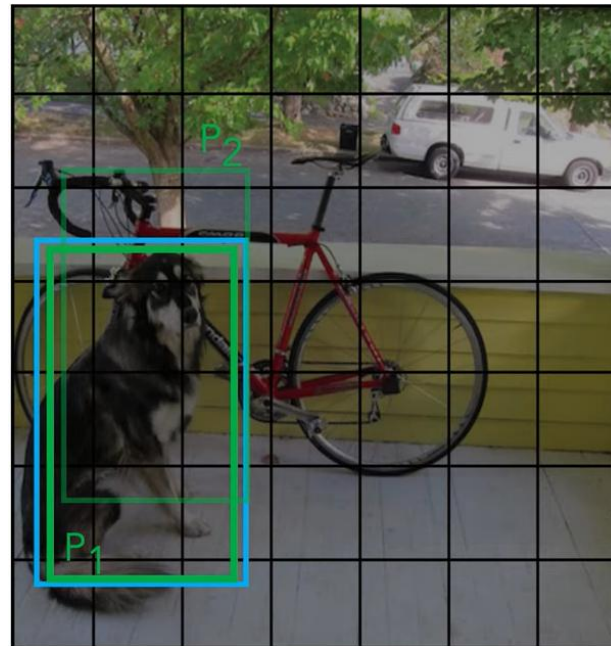
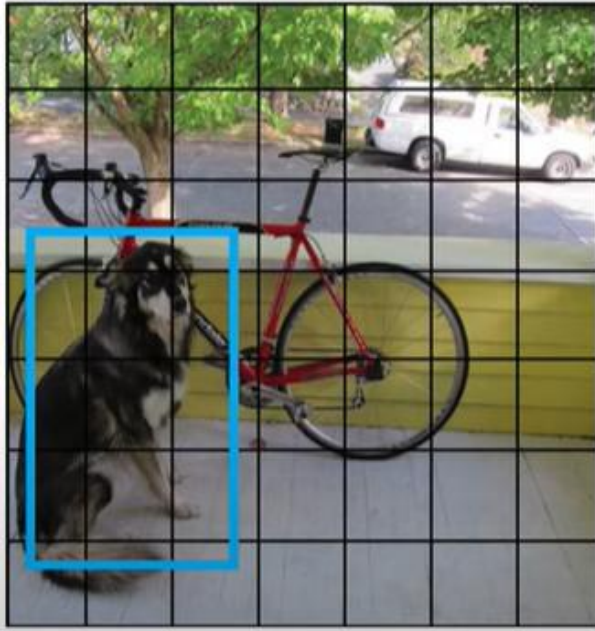
 = $IoU(pred, true)$



$$C = IoU(pred, true)$$

How does YOLO Work?

Box selection (Inference)



Let's predict this bounding box

If $IoU(P_1, P_2) > Threshold$:

$$P = \operatorname{argmax}(C(P_1), C(P_2))$$

How does YOLO Work?

Helps to put more weight to
box coordinates

Bounding box coordinate loss (Coordinate regression loss)

$$L = \lambda_{coord} \times \sum_{i=1}^{S^2} 1_i^{obj} \times \left((\Delta x_i^* - \Delta \hat{x}_i)^2 + (\Delta y_i^* - \Delta \hat{y}_i)^2 + \right. \\ \left. \left(\sqrt{\Delta w_i^*} - \sqrt{\Delta \hat{w}_i} \right)^2 + \left(\sqrt{\Delta h_i^*} - \sqrt{\Delta \hat{h}_i} \right)^2 \right)$$

If there is an object,
it is evaluated

$$+ \sum_{i=1}^{S^2} 1_i^{obj} \times (c_i^* - \hat{c}_i)^2 + \sum_{i=1}^{S^2} 1_i^{obj} \times \sum_{c=1}^{20} (p_{i,c} - \hat{p}_{i,c})^2$$

Confidence score loss

Class probability loss

$$+ \lambda_{no_obj} \sum_{i=1}^{S^2} 1_i^{no_obj} \times \sum_{j=1}^B (c_{i,j} - \hat{c}_{i,j})^2$$

If there is no object,
it is evaluated

For training part,
Loss function