

WEEK - 15

Aim :-

To implement a YOLO (You only look once) deep learning model for real-time object detection in images & videos.

objective :-

- To understand the concept of object detection using deep learning.
- To learn how YOLO divides images into grids and detects multiple objects at once.
- To implement a pre-trained YOLO v5 or YOLO v8 model using open source frameworks.
- To visualize bounding boxes and class labels on detected objects.

Algorithm :-

- 1.) Load a pre-trained YOLO model.
- 2.) Load an input image or video frame.
- 3.) Resize and Normalize image for model Input.
- 4.) Pass image through YOLO Network.
- 5.) Get predictions → bounding boxes, confidence scores, and class labels.
- 6.) Apply Non-Maximum Suppression (NMS) to remove duplicate boxes.
- 7.) Draw boxes and labels on the original image.
- 8.) Display detected objects.

Pseudo code :-

Import YOLO Library

Load pre-trained YOLO model and weights

Read Input Image or video

Resize and Normalize Image

Architecture:

Yolo v3 has 3 stages of detection
Stage 1: 320x320 grid, 32 classes, 53.4 mAP
Stage 2: 160x160 grid, 90 classes, 45.1 mAP
Stage 3: 80x80 grid, 90 classes, 41.5 mAP

Yolo v3 uses 3 different detection heads for different scales.

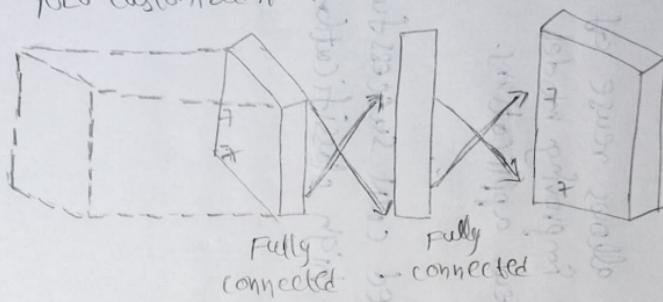
Detections for each class

Non-max suppression → Non-max suppression

YOLO Architecture

YOLO architecture consists of 3 stages of detection and 3 different output heads for different scales.

YOLO customized Architecture



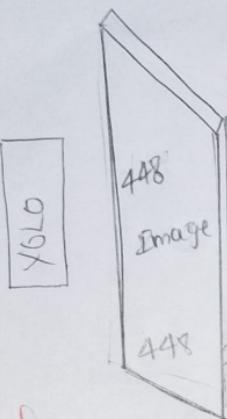
Fully connected

Fully connected

Fully connected

Input:

YOLO Image



output: 448 x 640 (no detections), 68.1 ms

speed : 10.7 ms preprocess, 68.1 ms Inference,

0.8ms postprocess per Image at shape (53, 448, 3)

Results ?
object confidence and bounding box

object	confidence	bnd box
person	0.93	Green box
car	0.89	Blue Box
Dog	0.85	Red Box

object detected : person (0.93)

object detected : car (0.89)

object detected : Dog (0.77)

Forward pass through YOLO model

For each detection :-

Extract confidence score and class id

If confidence > threshold :

compute bounding box coordinates

Apply Non-maximum suppression

Draw rectangle and label on image

Display output image with detections.

Observation :-

- YOLO detects multiple objects in a single pass efficiently.
- Each object has a bounding box, label and confidence score.
- Accuracy increases when lighting and angles are clear.
- Real-time detection works at high FPS.

Result :-

The YOLO model successfully detected and labeled multiple objects.

~~CHEM~~