



STAATLICH
ANERKANNTE
HOCHSCHULE

Data Analytics 4 - Project

Yolo - algorithm with use case

Presented on 10.06.2020 by:

Rohit Keshav Bewoor (11011831)

Big Data and Business Analytics 2018-20 batch
SRH Hochschule Heidelberg

Content

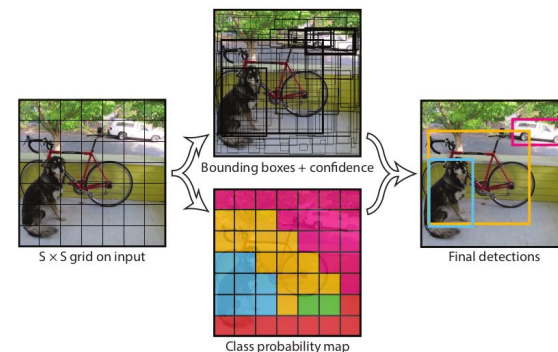
- Introduction
- How YOLO works
- Types of YOLO
- Use case
- Model Description and Visualisation
- References
- Q&A

Introduction

- You Only Look Once (YOLO) is an object detection algorithm
- Four versions - v1 in May 2016, v2 in December 2016, v3 in April 2018, v4 in April 2020
- Does not use region proposal based approach like the R-CNN family
- Developed using the Darknet framework
- Github Repo for this project work : https://github.com/rbewoor/DataAnalytics4_Project

How YOLO works

- Divides image into a grid of size $S \times S$ where S is an integer
- Each pixel evaluated as possible center point of an object
- All detections are evaluated in one pass - very fast algorithm
- Model is trained to identify C classes of objects
- B is the number of Bounding Boxes detected all over the image (without threshold consideration). Five values are output for each bounding box:
 - Two values for center coordinates
 - Two values for dimensions (height and width)
 - Confidence score
- Can handle multiple bounding boxes and aspect ratios (anchor box concept)
 - Anchor boxes are predefined boxes provided by the user to Darknet which gives the network an idea about the relative position and dimensions of the objects to be detected.
 - These are calculated using the training set Objects.



Source: YOLO v1 paper

How YOLO works

- Usually Non-max suppression used to remove redundant detections
- Total detections per image = $(S \times S) * (B * 5) + C$
- For example, suppose that:
 - image is divided into 3 x 3 grid (i.e. $S = 3$)
 - we want to detect dog, cat and bird (i.e. $C = 3$)
 - 10 boxes predicted ($B = 10$)
 - $\#Detections = (3 \times 3) * (10 * 5) + 3 = 477$
- Threshold value used for Confidence Score to evaluate acceptance of object detection

Types of Yolo

Model Type	YOLO v1	YOLO v2 (aka YOLO9000)	YOLOv3
Salient points	26 total (24 Conv + 2 FC) Problem detecting small objects	30 layers (included batch norm after every Conv) Anchor boxes introduced No FC present Still poor with small objects	106 layers Detection on 3 scales to handle small to large object sizes 9 anchor boxes (3 per scale)

FC: Fully connected layer

Conv: Convolution layer

YOLOv4 is very recent and not been studied in depth

Objective:

Use the YOLO model for desired use case and explain the architecture

Use Case:

Present a set of new images to a pre-trained YOLO v3 model.

For each image, capture the detected **object class** and the **confidence score**.

Store information in a neo4j graph database:

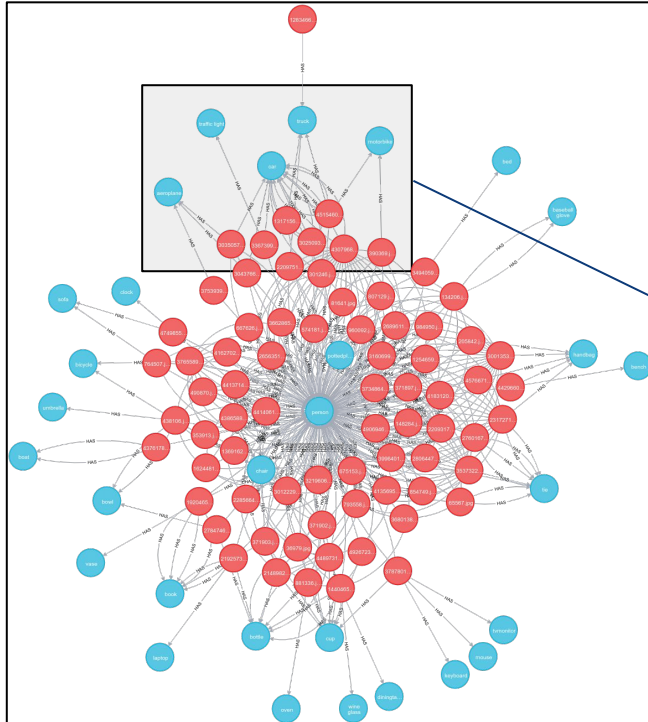
- Relationship format: (i:Image)-[r:HAS]->(o:Object)
- Confidence score is a property of the "HAS" relationship
- Python script: my_yolo3_one_file_to_detect_them_all_6.py

E.g. Image123.jpg HAS the objects:

- car (score 58.98),
- person (score 98.34)
- person (score 93.23)

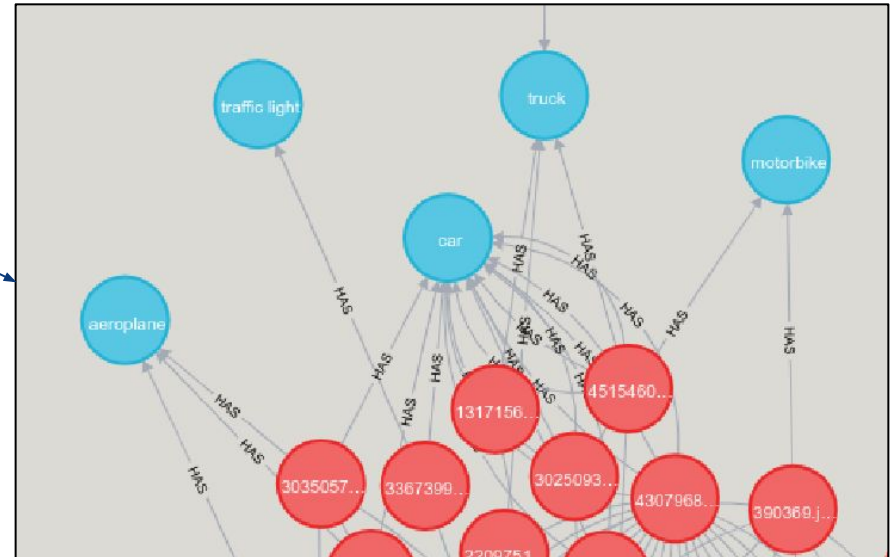
Neo4j representation

Image - HAS -> Object



Objects found in images: Traffic light, truck, motorbike, car, etc.

Many to many relationship could exist.



Model description and visualisation

- Darknet-53 block diagram taken from the YOLO v3 paper:
 - maps to Layers 0 - 74 of the source code
- Model description using Keras built in functions:
 - Python script: my_yolo3_model_stats_1.py
 - Textual description with model.summary function
Link: https://github.com/rbewoor/DataAnalytics4_Project/blob/master/model_summary_1.txt
 - Visualisation with Plot_model function
Link: https://github.com/rbewoor/DataAnalytics4_Project/blob/master/model_vis_1.png

	Type	Filters	Size	Output
1x	Convolutional	32	3 × 3	256 × 256
	Convolutional	64	3 × 3 / 2	128 × 128
	Convolutional	32	1 × 1	
	Convolutional	64	3 × 3	128 × 128
2x	Residual			128 × 128
	Convolutional	128	3 × 3 / 2	64 × 64
	Convolutional	64	1 × 1	
	Convolutional	128	3 × 3	64 × 64
8x	Residual			64 × 64
	Convolutional	256	3 × 3 / 2	32 × 32
	Convolutional	128	1 × 1	
	Convolutional	256	3 × 3	32 × 32
8x	Residual			32 × 32
	Convolutional	512	3 × 3 / 2	16 × 16
	Convolutional	256	1 × 1	
	Convolutional	512	3 × 3	16 × 16
4x	Residual			16 × 16
	Convolutional	1024	3 × 3 / 2	8 × 8
	Convolutional	512	1 × 1	
	Convolutional	1024	3 × 3	8 × 8
	Residual			8 × 8

References

1. How to Perform Object Detection With YOLOv3 in Keras.
<https://machinelearningmastery.com/how-to-perform-object-detection-with-yolov3-in-keras/>
2. Github Code <https://github.com/rbewoor/keras-yolo3> (forked from
<https://machinelearningmastery.com/how-to-perform-object-detection-with-yolov3-in-keras/> on 05.06.2020)
3. YOLOv3 pre-trained weights. <https://pireddie.com/media/files/yolov3.weights>
4. J. Redmon et al. You Only Look Once: Unified, Real-Time Object Detection. 09-05-2016. <https://arxiv.org/abs/1506.02640>
5. J. Redmon et al. YOLO9000: Better, Faster, Stronger. 25-12-2016. <https://arxiv.org/abs/1612.08242>
6. J. Redmon et al. YOLOv3: An Incremental Improvement. 08-04-2018. <https://arxiv.org/abs/1804.02767>
7. How to Visualize a Deep Learning Neural Network Model in Keras.
<https://machinelearningmastery.com/visualize-deep-learning-neural-network-model-keras/>
8. All About YOLO Object Detection and its 3 versions (Paper Summary and Codes).
<https://medium.com/data-science-in-your-pocket/all-about-yolo-object-detection-and-its-3-versions-paper-summary-and-codes-2742d24f56e>

Q&A

- Thank you.