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Final Project Group 7 Object Detection Using YOLO

Hao Heng, Ze Gong, Junhe Zhang

Introduction

- Darkent
- YOLO
- Create custom dataset
- Train YOLO network
- Result and Conclusion



What is darknet?

- Darknet is an open source neural network framework written in C and CUDA
- It is fast, easy to install, and supports CPU and GPU computation



What is Yolo?

 You only look once (YOLO) is a state-of-the-art, real-time object detection system



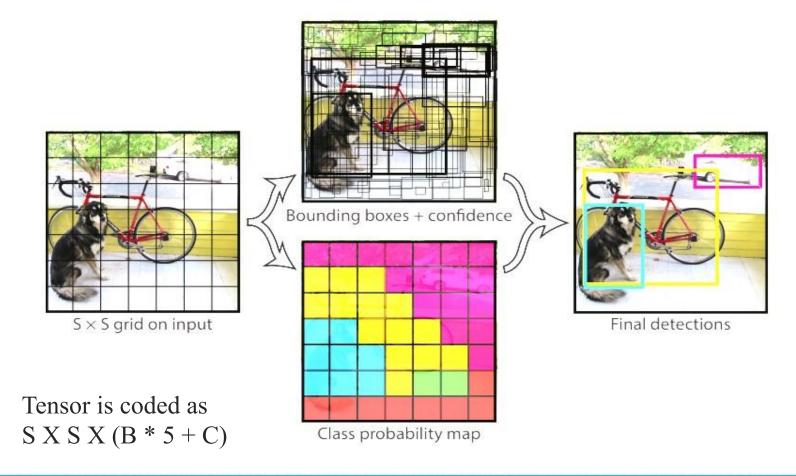
YOLO Network

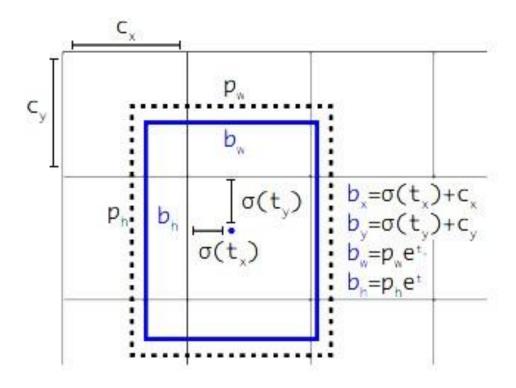
| | Type | Filters | Size | Output |
|----|---------------|---------|------------------|----------------|
| | Convolutional | 32 | 3 × 3 | 256 × 256 |
| | Convolutional | 64 | $3 \times 3/2$ | 128 × 128 |
| | Convolutional | 32 | 1 × 1 | |
| 1× | Convolutional | 64 | 3×3 | |
| | Residual | | No market water | 128 × 128 |
| | Convolutional | 128 | $3 \times 3 / 2$ | 64 × 64 |
| | Convolutional | 64 | 1 × 1 | |
| 2× | Convolutional | 128 | 3×3 | |
| | Residual | | | 64×64 |
| | Convolutional | 256 | 3 × 3 / 2 | 32 × 32 |
| | Convolutional | 128 | 1 × 1 | |
| 8× | Convolutional | 256 | 3×3 | |
| | Residual | | | 32×32 |
| | Convolutional | 512 | $3 \times 3 / 2$ | 16 × 16 |
| | Convolutional | 256 | 1 × 1 | |
| 8× | Convolutional | 512 | 3×3 | |
| | Residual | | | 16 × 16 |
| | Convolutional | 1024 | 3×3/2 | 8 × 8 |
| | Convolutional | 512 | 1 × 1 | -12 |
| 4× | Convolutional | 1024 | 3×3 | |
| | Residual | | | 8 × 8 |
| | Avgpool | | Global | |
| | Connected | | 1000 | |
| | Softmax | | | |

In this project we use YOLO (You Only Look Once) v3.0 for the purpose of Object Detection.



YOLO





Calculation of the bounding box coordinates.

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Create custom dataset

Dataset

Source: Google Open Image Dataset

link: https://storage.googleapis.com/openimages/web/ind

ex.html

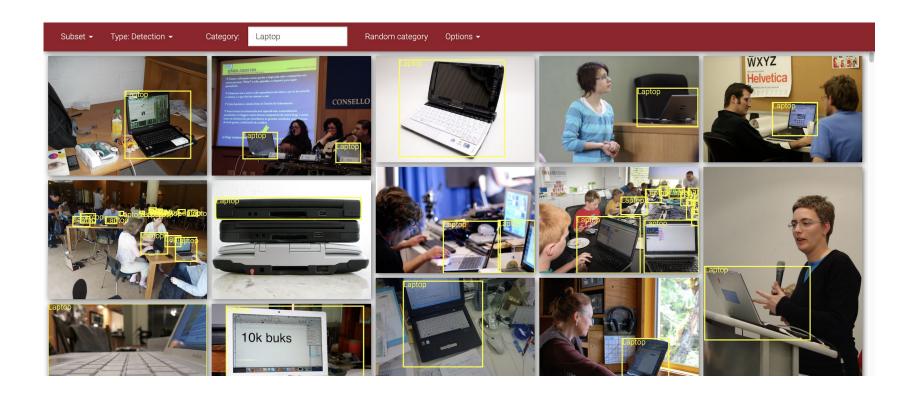
Downloader: OIDv4_ToolKit

Category: Car, Bicycle wheel, Traffic light, Person, Jeans, Laptop, Bus

Instance of each category: 1000



Dataset





Needed files for darknet framework:

- annotation for each image
- train.txt
- test.txt
- data.data
- classes.names



```
Annotation.txt (each image has its own .txt): class number center x center y width height
```



train.txt:

```
/full path/image1.jpg
/full path/image2.jpg
```

.

test.txt:

/full path/image1.jpg
/full path/image2.jpg

.



data.data:

classes=the number of classes in the dataset train=full path to train.txt file valid=full path to test.txt file names=full path to classes.names file backup=backup

classes.names:

Car

Bicycle wheel





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Training YOLO in Darknet Framework

Setting up Configuration

```
Classes = 7

filters = (7 + 5) x 3 = 36

max_batches = 7 x 2000 = 14000

steps = 0.8 x max_batches / 0.9 x max_batches

= 11200 / 12600

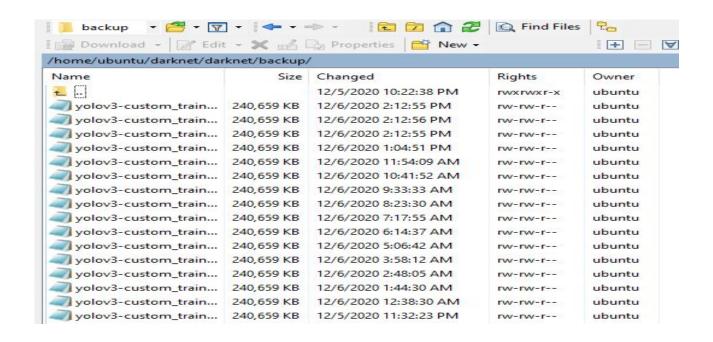
batch = 32

subdivisions = 8

minibatch = batch/subdivisions = 4
```



Result of Training Process

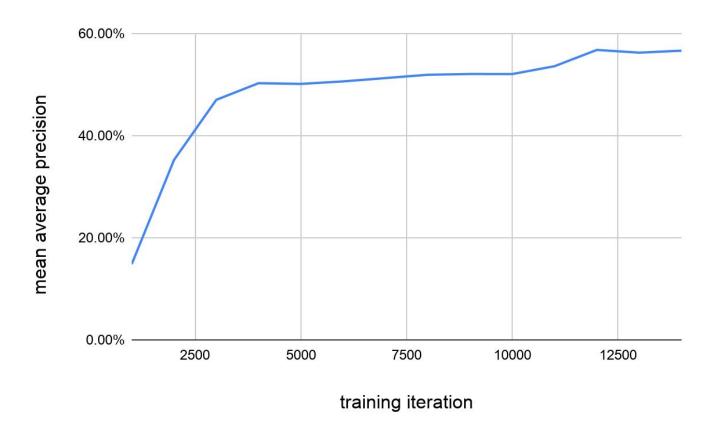


Mean Average Precision Example:

```
class id = 0, name = Car, ap = 51.01%
                                                (TP = 308, FP = 165)
class id = 1, name = Bicycle wheel, ap = 62.46%
                                                        (TP = 304, FP = 68)
                                                (TP = 161, FP = 35)
class id = 2, name = Bus, ap = 79.25%
class id = 3, name = Traffic light, ap = 44.14%
                                                      (TP = 329, FP = 222)
class id = 4, name = Jeans, ap = 55.24\%
                                                (TP = 183, FP = 64)
class id = 5, name = Laptop, ap = 80.80%
                                          (TP = 155, FP = 28)
class id = 6, name = Person, ap = 23.76%
                                                (TP = 230, FP = 308)
for conf thresh = 0.25, precision = 0.65, recall = 0.45, F1-score = 0.53
for conf thresh = 0.25, TP = 1670, FP = 890, FN = 2047, average IoU = 50.15 %
IoU threshold = 50 %, used Area-Under-Curve for each unique Recall
mean average precision (mAP@0.50) = 0.566669, or 56.67 %
```



Performance of Different Training Weights



The best mAP is 56.82%, appears at iteration 12000.



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Apply The Best Model on Testing image and video

Loading Best Model

There are two different ways to do object detection:

- Using darknet framework's command line
 - './darknet detector test cfg/custom_data.data cfg/yolov3-custom_train.cfg
 backup/yolov3-custom_train_last.weights data/laptop-jean-test.jpg -out_filename
 data/result-laptop-jean-test.jpg -dont_show'
- Run process_image_video.py file
 - python3 process_image_video.py



Algorithm of process_image_video.py

- Read the input image and convert it into a blob object which is accepted by the YOLO framework.
- 2. Load the best trained weights and the network structure file into the YOLO framework.
- 3. Process blob image file into the model and generate output bounding boxes for each object inside the image.
- Using Non-Maximun_Suppression technique on the output bounding boxes to drop duplicated boxes on the same object with lower confidence.



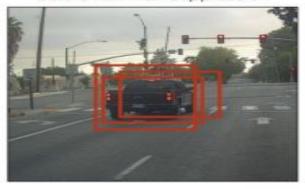
What the Results look like

```
out_scores, out_boxes, out_classes = predict(sess, "test.jpg")
Found 7 boxes for test.jpg
car 0.60 (925, 285) (1045, 374)
car 0.66 (706, 279) (786, 350)
bus 0.67 (5, 266) (220, 407)
car 0.70 (947, 324) (1280, 705)
car 0.74 (159, 303) (346, 440)
car 0.80 (761, 282) (942, 412)
car 0.89 (367, 300) (745, 648)
 100
 200
 300
 400
 500
 600
 700
```



Non-Maximum-Suppression

Before non-max suppression

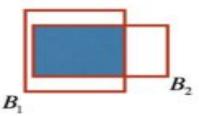


Non-Max Suppression

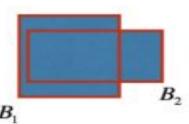
After non-max suppression



Intersection



Union



Intersection over Union

$$IoU = \frac{B_1 \cap B_2}{B_1 \cup B_2} = \frac{\Box}{\Box}$$



Results of Object Detected Images





Original Image

Model Detection



Results of Object Detected Images

Original Image





Model Detection



Results of Object Detected Images



Original Image



Model Detection



Results of Object Detected Videos



https://drive.google.com/file/d/1luAp5ggEG8hvPBlzG8yjvvBXh1PIXClW/view?usp=sharing





https://drive.google.com/file/d/1xc2RdESogHPag3VwwnbuNQs-RX--mZzF/view?usp=sharing



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Limitation and Conclusion

Limitations to YOLO

In spite of the high accuracy numbers and almost perfect bounding boxes, there are a few limitations to this object detection methods:

- Sometimes fails to detect overlapping objects, and objects that are partially visible in the frame.
- Detects the object class falsely if the features are a little blurred.
- It is very sensitive to model overfitting.



Conclusion and discussion

- The number of iteration of training process is not long enough to predict all categories we choose. (2000 iteration for each class)
- Our model is under-trained, compared to coco weight. Our dataset is not big enough to train a very accurate model
- Our model performs not well at detecting bus and person
- Our model has some difficult to detect object when there are multiple classes are overlapped. It will only detect one class.



References

Joseph Redmon, Santosh Divvala, Ross Girshick, ALi Farhadi "You Only Look Once: Unified, Real-Time Object Detection", IEEE Conference on Computer Vision and Pattern Recognition (CVPR), 2016.

Google open image: https://storage.googleapis.com/openimages/web/index.html

