

Regional Image Recognition using R-CNN

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Abstract

Image Recognition has, in recent history been relatively "solved" in the sense that algorithms have now been recorded outperforming humans in simple image recognition tasks. However regional image recognition where the algorithm is tasked with recognizing multiple images inside a whole chaotic scene is still an emerging field. Our team analysed the most promising methods of regional image recognition, and implemented our own solution to a simplified regional recognition task. Our solution *was able to locate and classify objects inside an image with a 92% accuracy of locations and a 100% accuracy of classification.*

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Contents

1	Introduction	3
1.1	Data sets	3
1.2	Our Aims	3
2	Design	3
2.1	Regional Convolutional Neural Networks (RCNNs)	4
2.2	"You Only Look Once" (YOLO)	4
2.3	Selective Search	4
3	Implementation	4
4	Experiments	4
5	Conclusion	4
6	Description of Collaboration	4

1 Introduction

For this project, our team was assigned the task of implementing a regional image recognition system.

1.1 Data sets

The data set provided was given in the following format:

- Data: 400x400 RGB image files in .jpg format
- Labels: each image has a corresponding text file that describes the location of each of the predefined objects in the image. This location was given as pairs of integers describing horizontal runs of pixels that form a rectangular bounding box around the object. If the object was not present in the image it was given as [object 1 0] meaning that it had a run length of 0 pixels.

Training Set The training set contained n (image, label) pairs

Test Set The test set contained m (image, label) pairs

1.2 Our Aims

Our aims for this project are as follows:

- Discover and Analyse currently existing method of performing regional image recognition
- Produce our own implementation of one of these methods, using any necessary packages or source code segments as necessary.
- Conduct experiments to optimize the recognition system in terms of bounding box locations, and object classification.
- Produce a set of conclusions about the effectiveness of the various recognition methods, and the optimal parameters of our implementation for the data set provided.

2 Design

–Options available to us:

- RCNN, Fast, Faster
- Yolo
- Selective Search
- general backprop

2.1 Regional Convolutional Neural Networks (RCNNs)

Given the regional nature of this task, The first option that should be analysed is "Regional Convolutional Neural Networks" and their successors. There are three implementations of this algorithm that will be examined:

- R-CNN
- Fast R-CNN [Gir15]
- Faster R-CNN [RHGS15]

R-CNN R-CNN works by....

Fast R-CNN Fast R-CNN [Gir15] is an update on the original R-CNN technique that was developed in 2015, it achieves approximately a 10x speed-up on the original. It does this by....

Faster R-CNN Faster R-CNN [RHGS15] is another significant update on the Fast R-CNN technique that achieves another dramatic speedup. This algorithm was designed as part of an attempt at real time regional image recognition, and as such is able to operate in almost real time.

2.2 "You Only Look Once" (YOLO)

"You Only Look Once" (YOLO) was named as such because the theory centers around only performing a single pass across the image, rather than having to analyse the same data multiple times.

2.3 Selective Search

Selective Search is an algorithm used for regional image searching....

3 Implementation

-Using RCNN -TensorFlow Layers

4 Experiments

5 Conclusion

6 Description of Collaboration

Everyone did good work. brendan and bill - RCNN Priyanka Liangye - YOLO

References

- [Gir15] Ross Girshick. Fast r-cnn. 2015.
- [RHGS15] Shaoqing Ren, Kaiming He, Ross Girshick, and Jian Sun. Faster r-cnn: Towards real-time object detection with region proposal networks. 2015.