

Refined Methodology for Accurately Detecting Objects from Digitized Herbarium Specimen

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MAMUDS
MANAGING MULTIMEDIA DATA FOR SCIENCE

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#1 BACKGROUND

- Huge amount of digitized herbarium specimens.
- Providing valuable information on functional traits.
- Deploying computer vision techniques to automatically extract functional trait data from digitized specimen images.
- Recognizing objects present on the specimen images which may appear in any positions and have different sizes (Figure 1).
- Developing a refined methodology RefYOLO, which is based on the pioneer object recognition system called You Only Look Once (YOLO).
- RefYOLO is darknet based framework for recognizing objects from digitized herbarium specimens.
- RefYOLO is one of the main outcomes of the Managing Multimedia Data for Science (MAMUDS) project.

#2 OBJECTIVES

- Recognizing efficiently all object classes of specimen images in order to extract automatically measurements of plant leaves.
- Batch processing of large specimen image data sets.

#3 ACHIEVEMENTS

- Achieve an accuracy better than two-stage object detectors (Figure 2) (Table 1).
- Achieve a fast recognition of objects within digitized herbarium specimens (0,06 s) (Table 2).
- Encoding contextual feature information about classes such as shape, contours and their appearance.
- Independent of object positions, orientation, size and shape.
- Recognize perfectly all object classes within the digitized herbarium specimens (7 classes).
- Optimizing complexity of the model in term of number of parameters.
- Introducing the Average-Max pooling method (AM-PM) to increase the recognition accuracy.

#4 SPECIMEN IMAGES



Figure 1. An example of a typical digitized herbarium specimens

#5 RefYOLO ARCHITECTURE

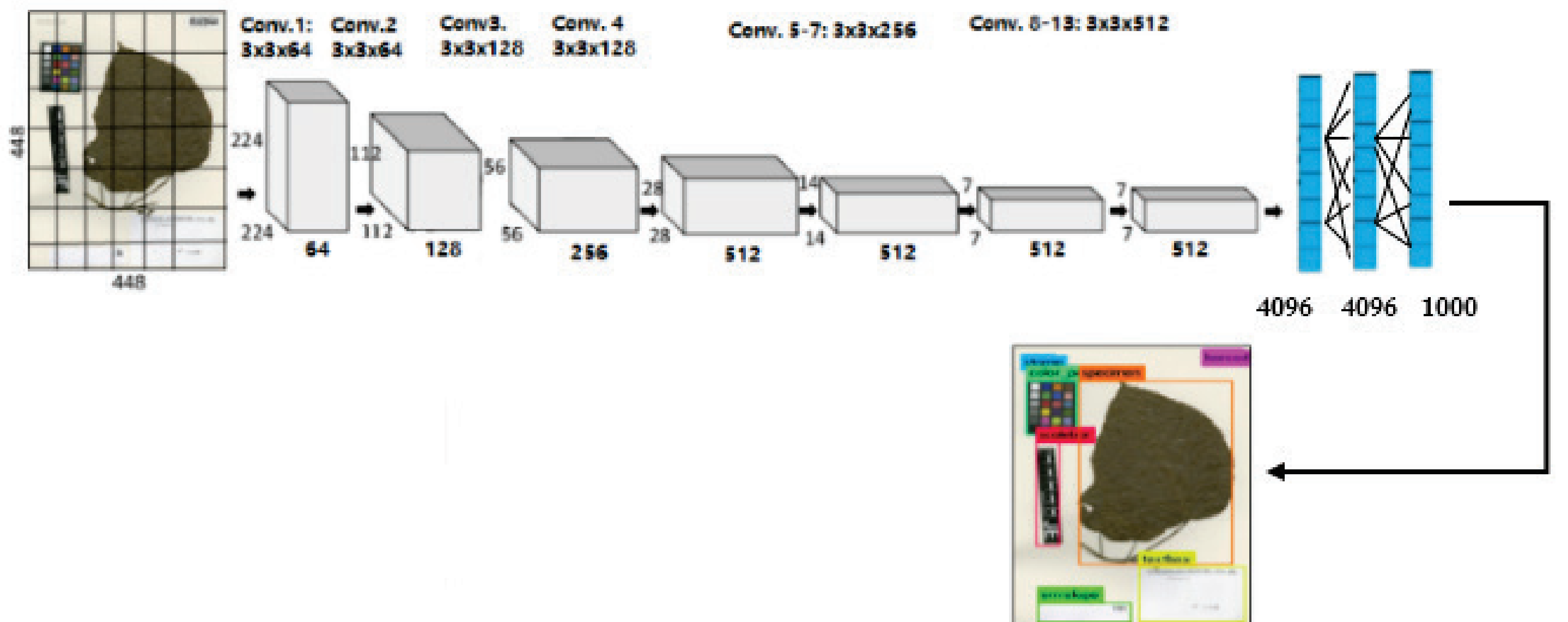


Figure 2. Overall Architecture

#6 EXPERIMENTS

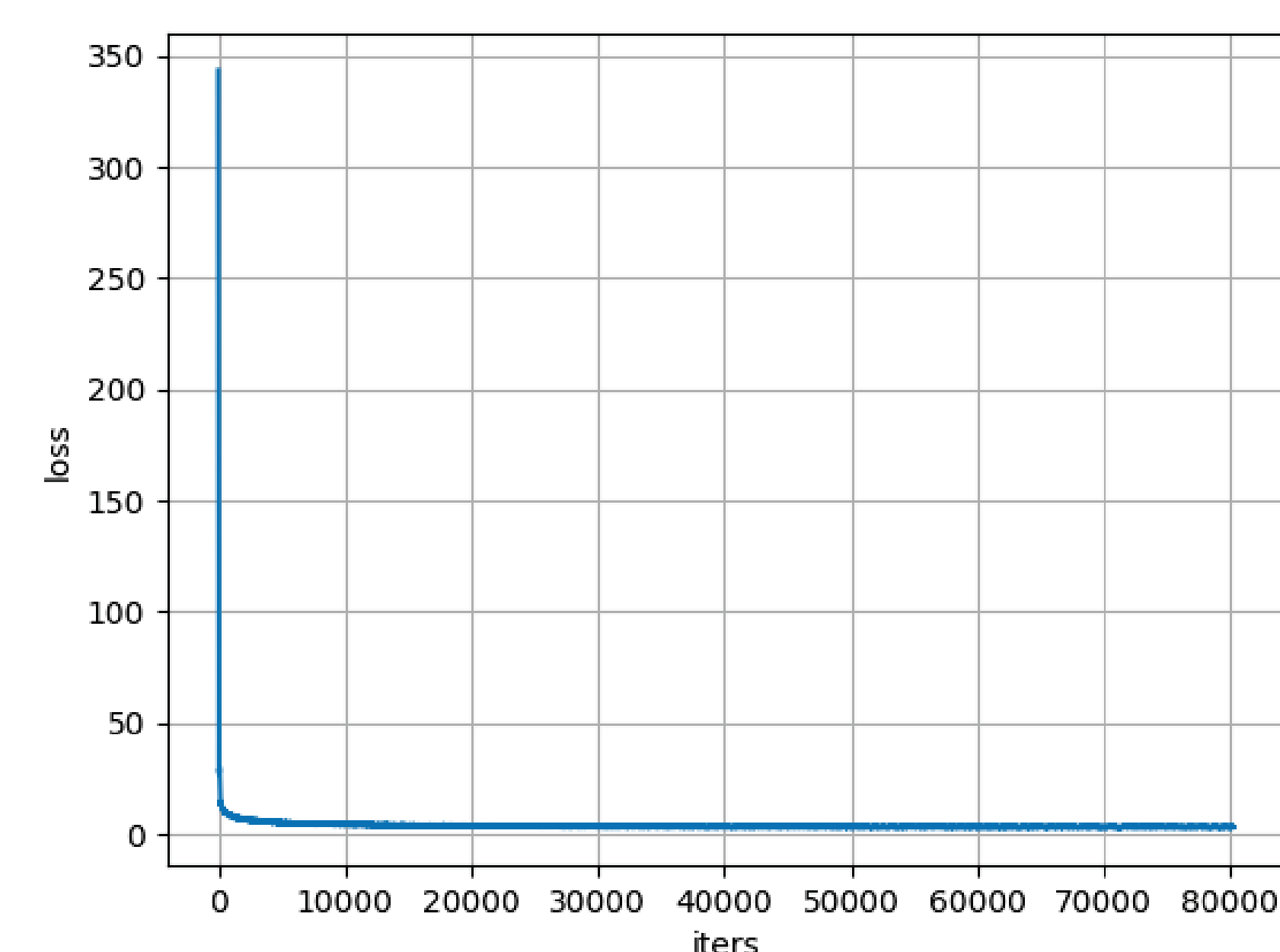


Figure 3. Iteration by loss graph

Table 1. mAP Measurements.

| | RefYOLO | YOLO | Faster RCNN | SSD | OverFeat |
|-----|---------|------|-------------|-----|----------|
| mAP | 89% | 77% | 84% | 76% | 79% |

Table 2. mIoU Measurements.

| Objects | RefYOLO | Faster RCNN | SSD | OverFeat |
|----------------|---------|-------------|-----|----------|
| Stamp | 80% | 84% | 76% | 71% |
| Barcode | 86% | 81% | 85% | 87% |
| ScaleBar | 88% | 87% | 80% | 78% |
| Color-Pallet | 92% | 90% | 87% | 70% |
| Specimen label | 91% | 78% | 82% | 81% |
| Envelope | 92% | 89% | 77% | 82% |
| Specimen | 94% | 90% | 88% | 79% |
| Time(s/image) | 0.06 | 0,14 | 0,2 | 0,6 |

#7 CONCLUSION and FURTHER WORK

- Proposed a new Deep Learning based Model for recognizing efficiently all object classes of specimen images
- Operate on batch processing of large specimen image data sets.
- Next step will be focused on fine segmentation of herb regions.