Object detection applied to the problem of robotic controlled pollination

Piyush Pandey 2024/04/19

Overview

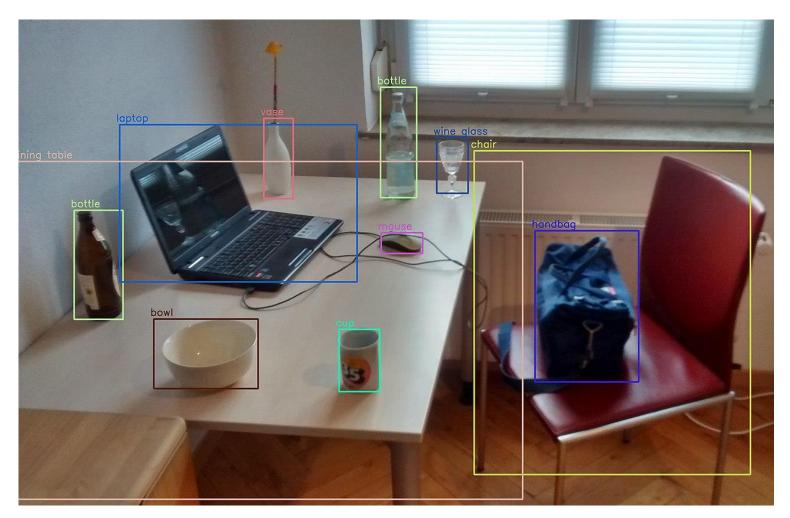
1. Object detection

2. Controlled pollination and the need for bag detection

3. Model architecture and workflow

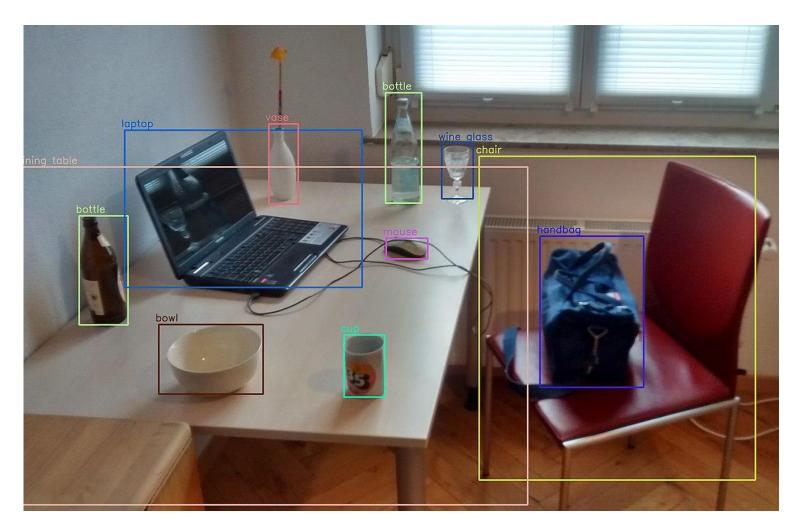
4. Code implementation (Google Colab)

Object detection



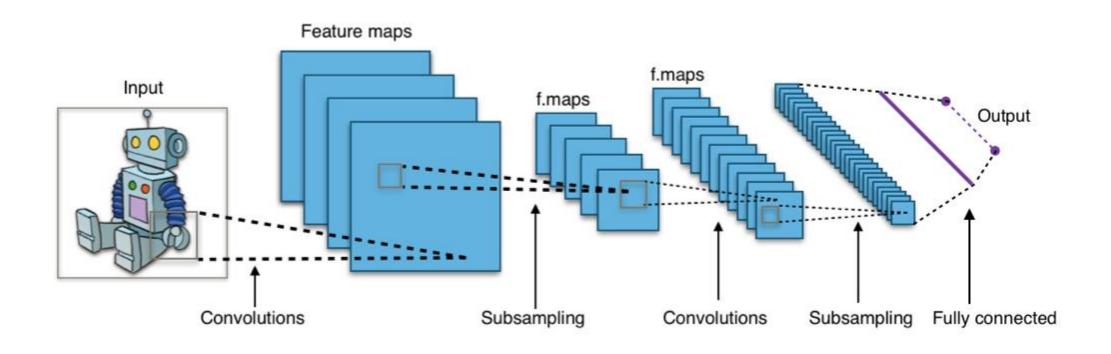
- Object detection includes the localization and classification of objects in the image
- Localization : bounding box
- Classification: box labels

Image classification



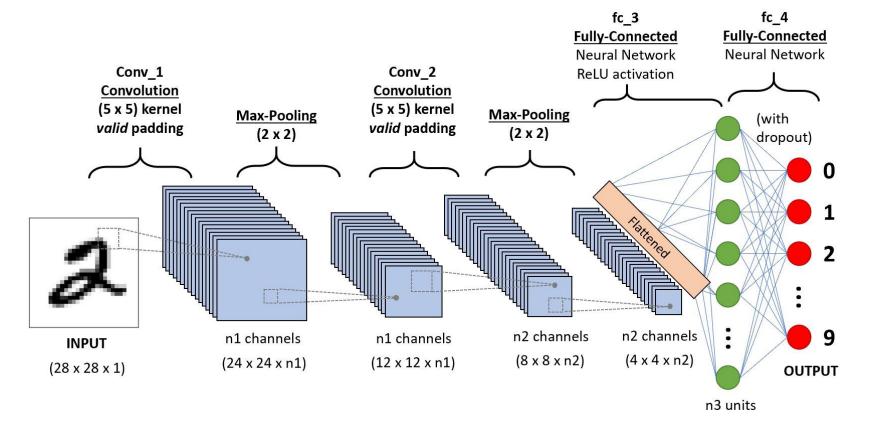
- Classification requires the assignment of a single label to the entire image
- A possible label for this image?

Image classification with CNNs



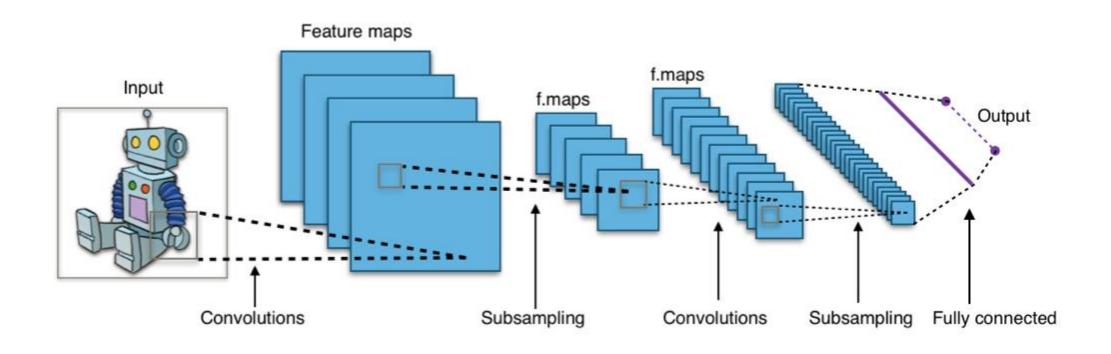
Typical CNN model with convolutional and subsampling (pooling) layers followed by fully connected layers

Image classification with CNNs



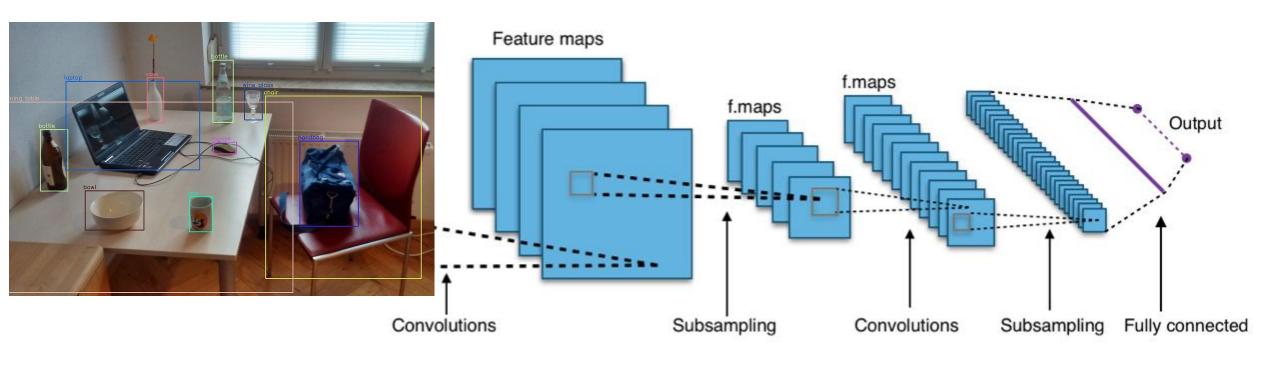
Typical CNN model with convolutional and subsampling (pooling) layers followed by fully connected layers

Object detection with CNNs



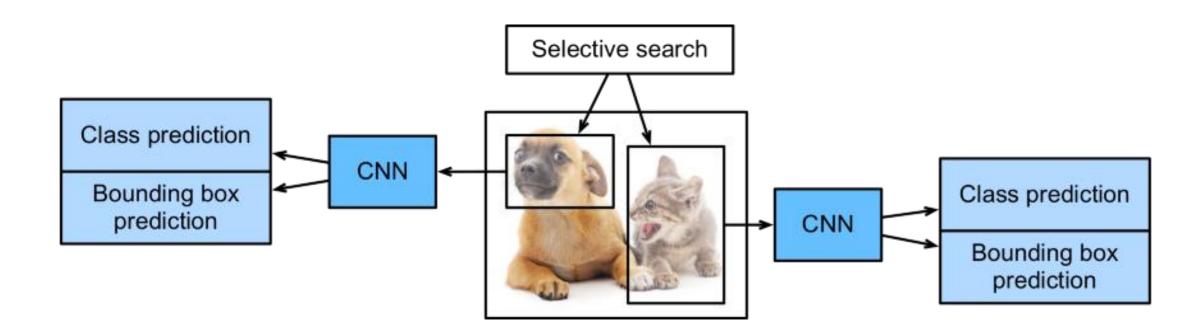
Object detection can be understood as the problem of operating this typical CNN model on an image patch

Object detection with CNNs

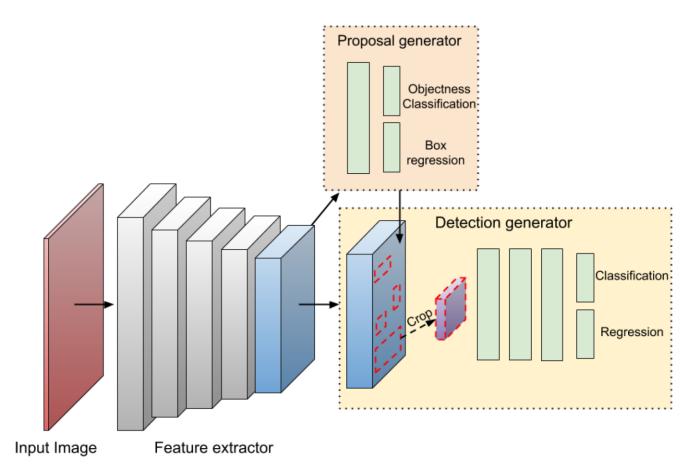


Object detection can be understood as the problem of operating this typical CNN model on an image patch

Object detection with CNNs



Two-stage object detectors



In a two-stage object detector, possible patches of object location are first identified followed by the classification of these patches and the refinement of the location

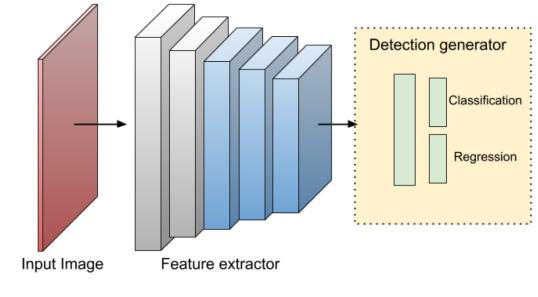
(a) Two-stage Faster R-CNN

Carranza-García, M., Torres-Mateo, J., Lara-Benítez, P., & García-Gutiérrez, J. (2020). On the performance of one-stage and two-stage object detectors in autonomous vehicles using camera data. *Remote Sensing*, 13(4), 89.

Two-stage object detectors

In a one-stage object detection model, bounding box location and label are simultaneously predicted by the model

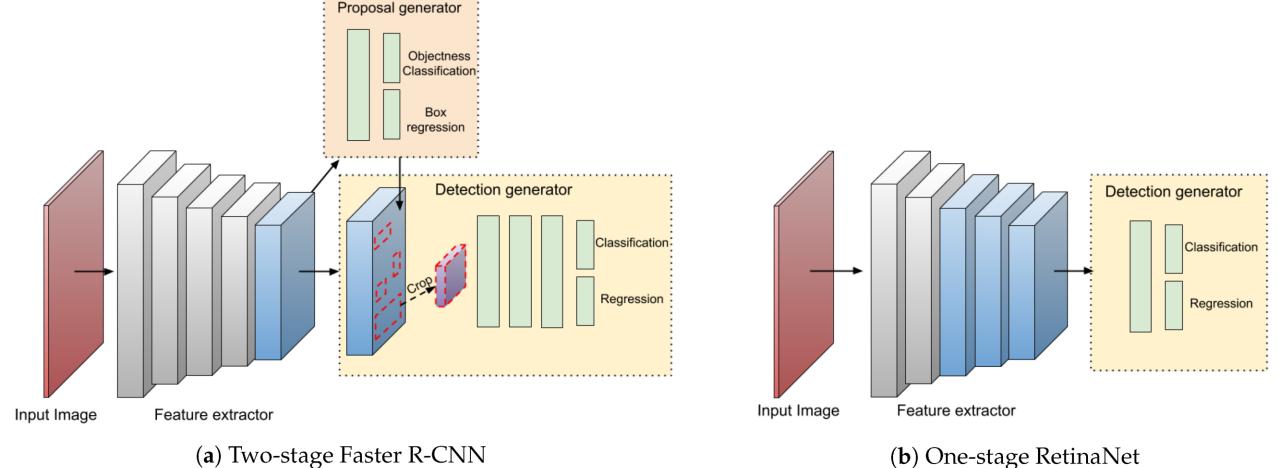
The recognition of the patch with possible object is included in the model architecture



(b) One-stage RetinaNet

Carranza-García, M., Torres-Mateo, J., Lara-Benítez, P., & García-Gutiérrez, J. (2020). On the performance of one-stage and two-stage object detectors in autonomous vehicles using camera data. *Remote Sensing*, 13(1), 89.

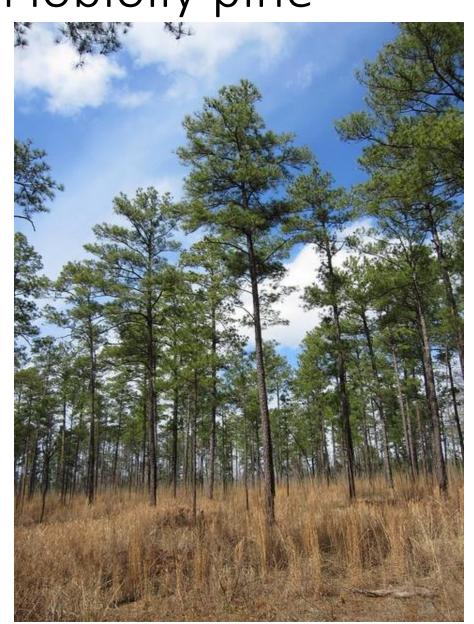
Two-stage object detectors



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- Timber tree used for pulp, plywood, general construction
- Dominant tree species in NC and the southeast

• In 2019, forest sector generated \$21.6 billion in industry output¹



¹ Parajuli, R.et al. (2020). *Economic Contribution of the Forest Sector in North Carolina, 2019.* NC State Extension, NC State University, College of Natural Resources.





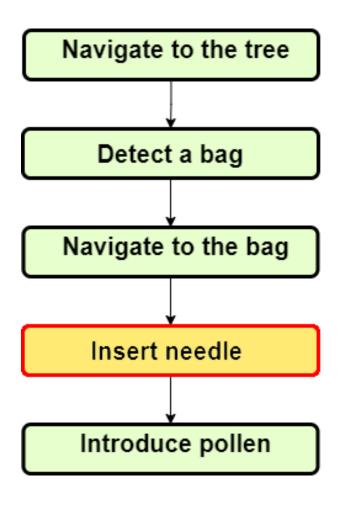






- More than a million bags placed annually
- Narrow time-window for pollination
- Challenges due to weather
- Safety
- Soil compaction



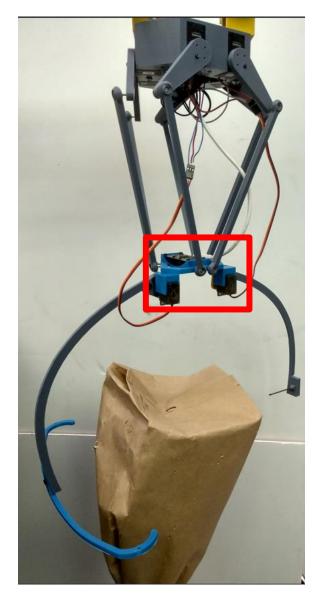






DUO MLX Stereo camera







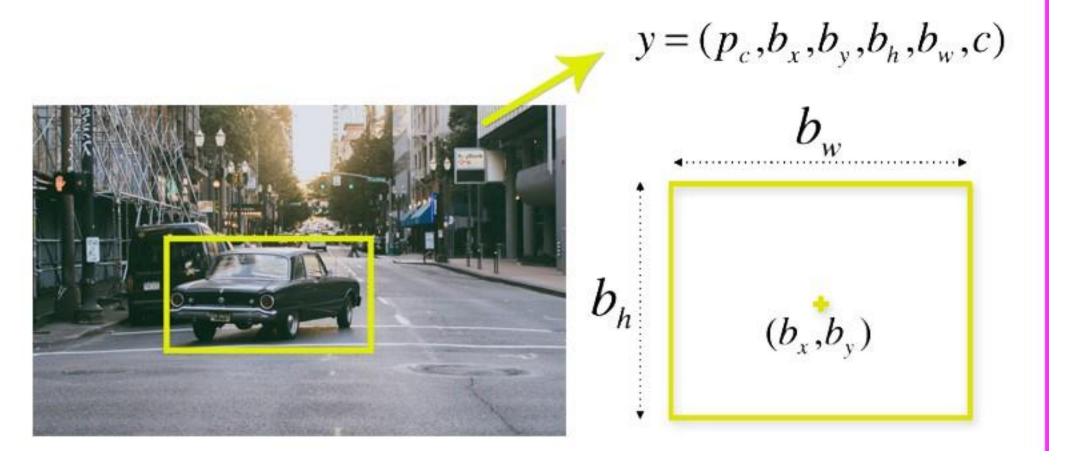
DUO MLX Stereo camera

Classification or detection?

One-stage object detection model or a two-stage model?

What about segmentation?

You Only Look Once (YOLO)



https://appsilon.com/object-detection-yolo-algorithm/

We predict bounding box location (b_x , b_y , b_h , b_w), probability of objectness (p_c) and the object class label (c).

You Only Look Once (YOLO)

preprocessed image encoding (608, 608, 3)(19, 19, 5, 85)19 Deep CNN 19 reduction factor: 32 $p_c b_x b_y b_h b_w$ 80 class probabilities Instead of finding "possible object locations", box 1 box 2 we process all locations box 3 box 4

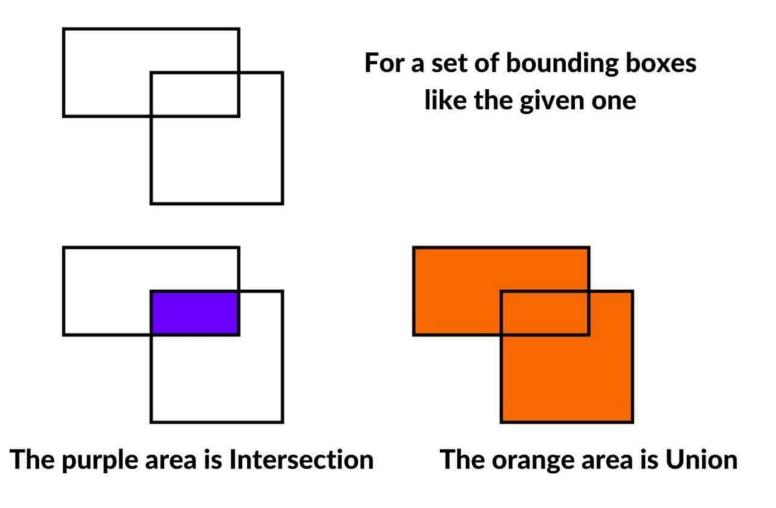
box 5

https://appsilon.com/object-detection-yolo-algorithm/

YOLO Idea 1: NMS



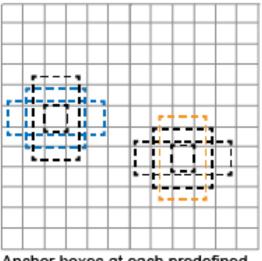
YOLO Idea 2: IoU



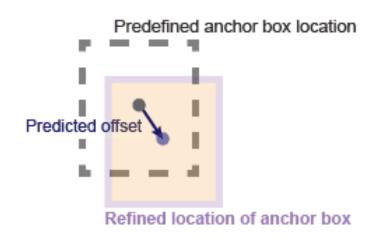
YOLO Idea 3: Anchor boxes

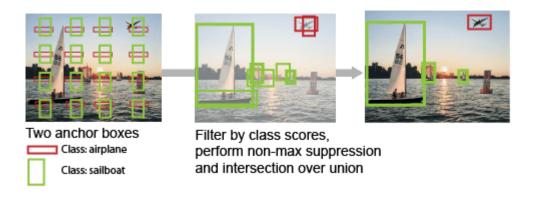


Ground truth image and bounding boxes



Anchor boxes at each predefined location in each feature map





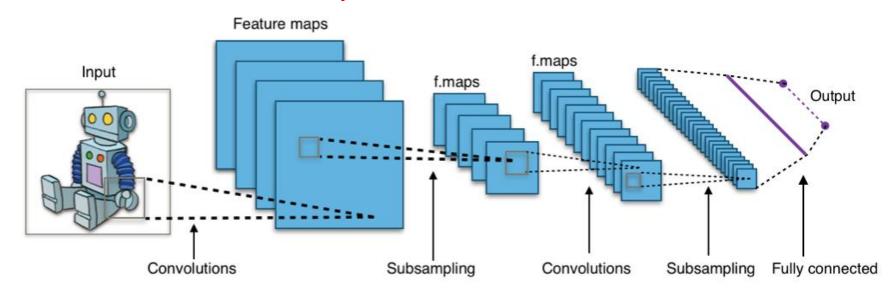
Model training pipeline

- 1. Select model architecture
- 2. Decide training process (Training from scratch? Transfer learning?)
- 3. Label images
- 4. Split data into train-validation-testing
- 5. Train the model and iterate

Transfer learning

The initial layers are frozen

Existing model architecture, pre-trained on large dataset...





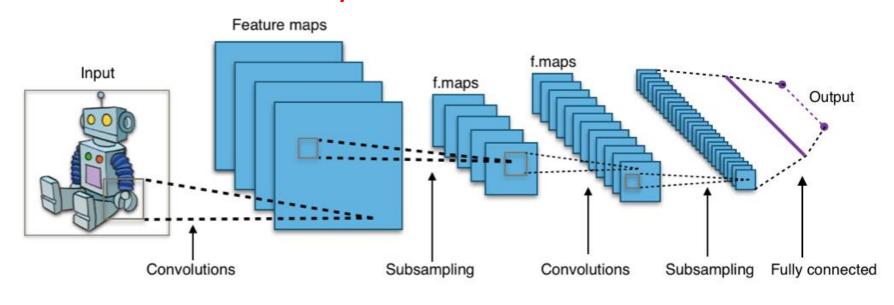
But not trained on paper bags put on tree branches.

Freezing parts of the pretrained model (keeping the trained parameters unchanged) leads to faster training on the novel dataset.

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Model training pipeline

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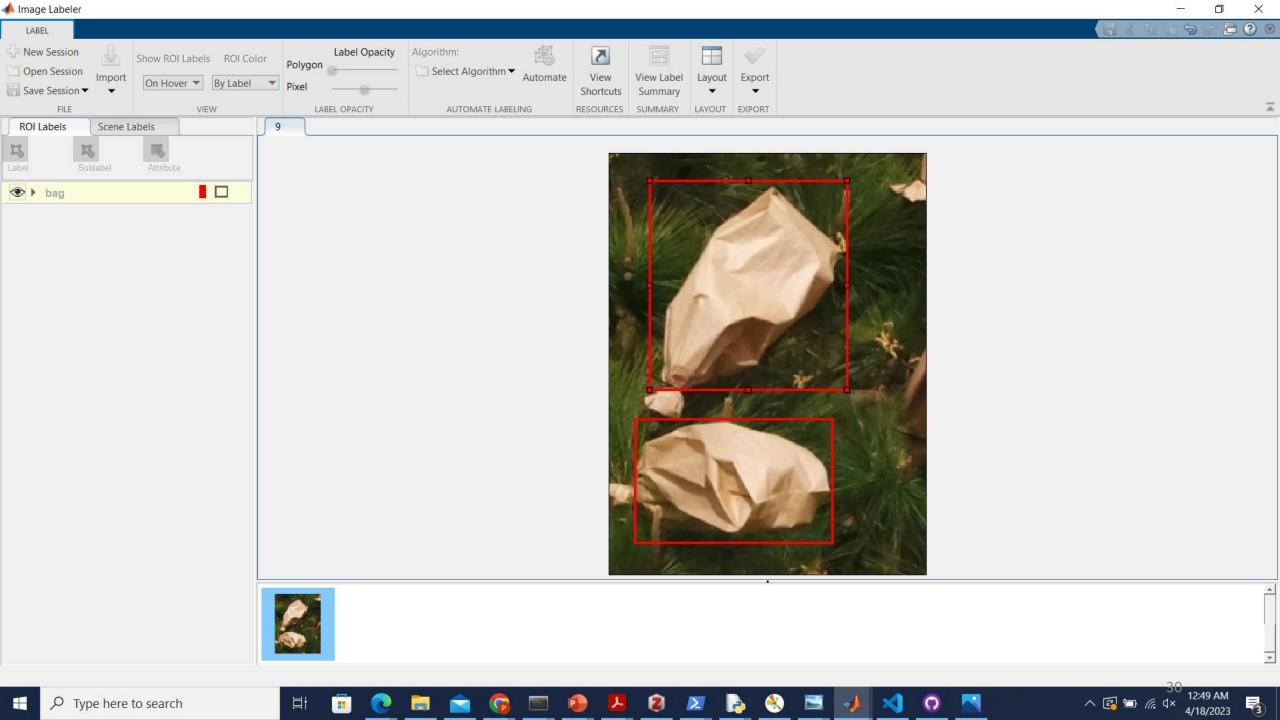
Labeling tools

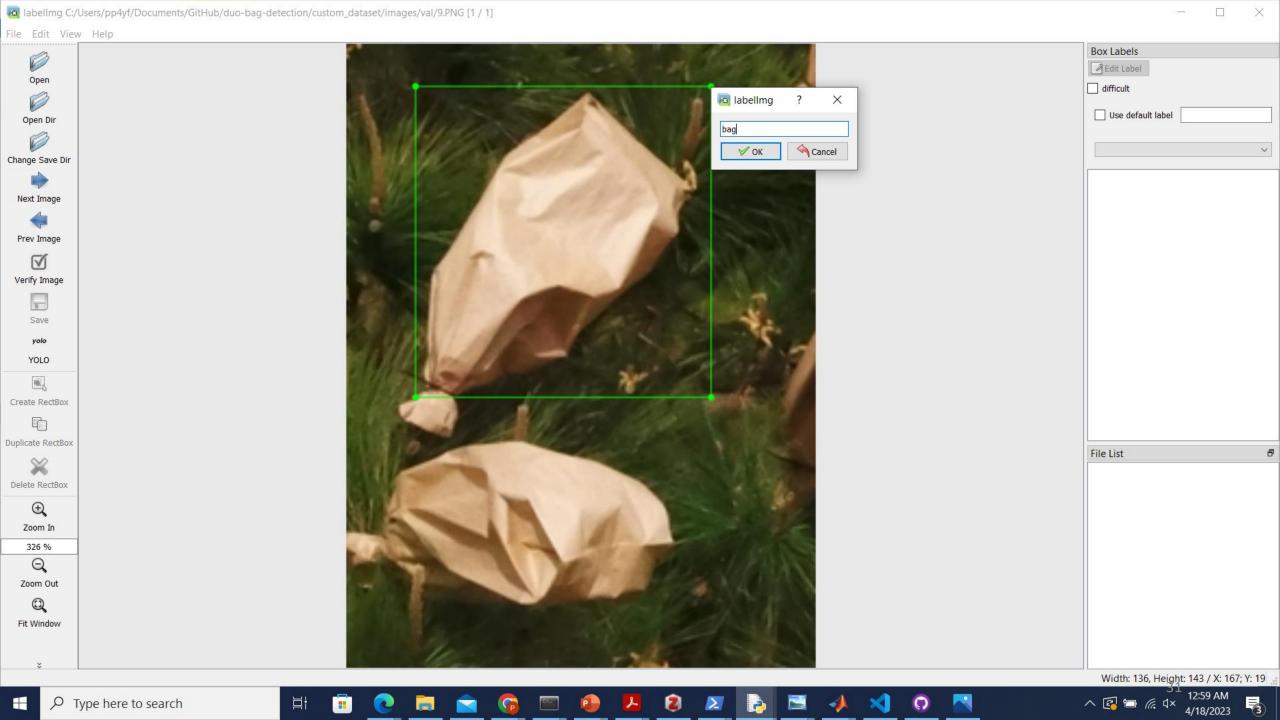
Labeling for object detection is simply drawing a rectangle and assigning a class label

- 1. MATLAB image labeler: Computer vision toolbox Stable, intuitive, and reliable, not free
- 2. LabelImg: https://github.com/heartexlabs/labelImg (Transitioned to a new tool (Label Studio))

Intuitive and free, limited functionality

3. VGG Image Annotator: https://www.robots.ox.ac.uk/~vgg/software/via/ Simple and free but not user friendly





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For steps 4 and 5, visit https://github.com/piyuss/bag-detection, click on Exclusion_bag_detection_demo_YOLOv5.ipynb, click on Open in Colab You will need a Google account and will need to log in to your account to run the code.