Machine Learning for Agricultural Applications

Assignment 8

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Object detection with YOLOv3

In this assignment you will work with a pre-trained model of YOLOv3. The original code of YOLO¹ (https://pjreddie.com/darknet/yolo/) is written in C, hence we use a convertion to TensorFlow/Keras from the website: https://machinelearningmastery.com/how-to-perform-object-detection-with-yolov3-in-keras/. Read the tutorial and follow the instructions starting below Create and Save Model.

You will load a model, that is already trained. That means, for single image processing you don't need a GPU, because it takes only a few seconds on a normal CPU. Therefore, you can run it either on your local installation or on Colab.

If you work on Colab, you need to load the YOLOv3.weights data set (237 MB) to Google Drive and mount the Google Drive in Colab:

```
from google.colab import drive
drive.mount("/content/gdrive", force_remount=True)
```

For reading the weights, you need to include a path for instance like '/content/gdrive/My Drive/Yolo/yolov3.weights'



Figure 1: Object detection with Yolo. Do you see the bus? ;-)

 $^{^1\}mathrm{You}$ Only Look Once: Unified, Real-Time Object Detection Joseph Redmon, Santosh Divvala, Ross Girshick, Ali Farhadi, CVPR 2016

Download the YOLOv3.weights to your computer or to Google Drive.

Use the make_yolov3_model() function to define you model. Read and load the weights, stored in the file yolov3.weights. Save the model in a model.h5-file.

How many convolutions are included in the model?

How many parameters are used in the model?

Task 2 - Load Model and Detect

[15 points]

Now make a new python script that loads the model and applies it to the set of images linked in moodle below the exercise sheet. Save the resulting images.

Print out the coordinates of the bounding boxes $(x_{\min}, y_{\min}, x_{\max}, y_{\max})$.

Why are some objects not identified or miss-classified?

Task 3 – Intersection over Union (IoU)

[15 points]

Write a python script, that calculates the Intersection over Union (IoU) of the ground throuth bounding box (yellow area in Figue 2) with coordinates $(x_{\min}, y_{\min}, x_{\max}, y_{\max}) = (360, 237, 2262, 1752)$ and the predicted bounding box (white frame in Figure 2). Use the coordinates of the predicted frame for img_7.jpg from Task 2.

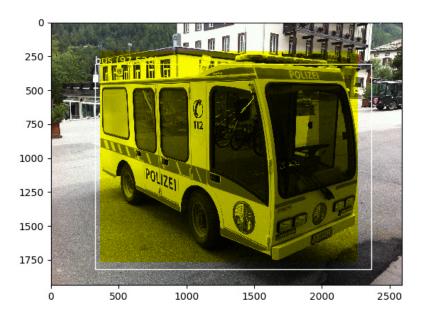


Figure 2: Ground trouth bounding box (yello area) and prediction (white frame).

```
TASK 1:
ex8_create_model.py
                      72 (?)
Convolutions:
Total params:
                      62,001,757
Trainable params:
                      61,949,149
Non-trainable params: 52,608
TASK 2:
ex8_load_model_detect.py
                             (VM_Linux auf meinem Laptop)
Bounding Box of img_7.jpg:
box.ymin, box.xmin, box.ymax, box.xmax:
                                          328 304 2364 1820
probability bus: 97.52770066261292
TASK 3:
ex8_IoU.py
# Area of both bounding boxes
A1 = np.abs(xmax_1 - xmin_1)*np.abs(ymax_1 - ymin_1)
A2 = np.abs(xmax_2 - xmin_2)*np.abs(ymax_2 - ymin_2)
# area of intersection:
A_inter = float(np.abs(xmax_1 - xmax_2)*np.abs(ymax_1 - ymax_2))
# union = A1 + A2 - A_inter
A_{union} = float(A1 + A2 - A_{inter})
IoU = np.abs(A_inter / A_union)
ground_trouth (xmin, ymin, xmax, ymax): 360 237 2262 1752
prediction
              (xmin, ymin, xmax, ymax):
                                          328 304 2364 1820
width ground truth:
                      1902
higth ground truth:
                      1515
width prediction :
                      2036
higth prediction :
                      1516
A1, A2:
                    2881530 3086576
A_inter, A_union:
                    6936.0 5961170.0
```

IoU: 0.0011635299781754254