

Configuration Manual

MSc Research Project
Programme Name

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Project Submission Sheet
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Configuration Manual

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1 Introduction

This project detects objects in traffic camera images using YOLOv3, a fast neural network for localization and recognition. The goal is to leverage a pre-trained YOLOv3 model to identify cars, pedestrians, cyclists, and other objects in the KITTI traffic camera dataset. KITTI contains real-world driving footage, providing diverse road scenes. Object detection has key applications in autonomous vehicles, traffic analysis, and safety systems. This project will run YOLOv3 inference on KITTI images, visualize detections, and evaluate performance. The results will demonstrate YOLOv3's ability to quickly and accurately detect objects of interest in traffic images. Overall, this project establishes a basis for deploying real-time object detection in traffic video analysis using a state-of-the-art deep learning model.

2 Prerequisites

2.1 Software

Python 3.x, OpenCV 4.x or later, Darknet framework, Additional Python packages like matplotlib, numpy etc.

2.2 Hardware

AMD EPYC Processors support for efficient model training and inference,

2.3 Dataset

KITTI object detection dataset images and annotations, Camera images containing road/traffic scenes, Annotation files with object bounding boxes, Classes: car, pedestrian, cyclist etc., Training and validation splits, Images should be extracted and preprocessed.

3 Environment Setup

3.1 Install Python Packages:

- Create and activate a Python 3.7+ virtual environment
- Install OpenCV using pip install opencv-python
- Install other packages like matplotlib, numpy, pandas etc.

3.2 Download and Compile Darknet

- Clone the Darknet repo: `git clone https://github.com/pjreddie/darknet`
- Enter the darknet directory and edit the Makefile to enable GPU and CUDA
- Build Darknet by running `make`

3.3 Get Pre-trained Weights and Config

- Download YOLOv3 weights file from GitHub
- Get `yolov3.cfg` configuration file from the darknet repository
- Save weights and `cfg` file in the darknet directory

3.4 Verify Installation

- Run darknet detector test to validate Darknet build
- Load a sample image and confirm YOLOv3 model detects objects

4 Implementation

4.1 Import Modules:

Import the necessary modules

```
import numpy as np
import time
import cv2
import sys
import os
import glob
```

```
data_path = '/content/drive/MyDrive/KittiDataSubSet'
yolo_path = '/content/drive/MyDrive/yolo-coco'
confidence_threshold = 0.5
nms_threshold = 0.3
limit = 421
```

```
images = glob.glob(data_path + '/*.png')
annotations = glob.glob(data_path + '/*.txt')
```

- Sets `data_path` and `yolo_path` variables to locate image dataset and YOLO model files.
- Defines confidence threshold of 0.5 and NMS threshold of 0.3 for model detection.
- Uses `glob` to get image and annotation file paths from `data_path`. Images and annotations will be used to run YOLOv3 model and evaluate detections.

Going back to the content folder.

```
!wget https://pjreddie.com/media/files/yolov3.weights
!wget https://raw.githubusercontent.com/pjreddie/darknet/master/cfg/yolov3.cfg
```

Python

Downloads YOLOv3 pre-trained weights and configuration file using wget.

```
2023-08-06 04:19:12-- https://pjreddie.com/media/files/yolov3.weights
Resolving pjreddie.com (pjreddie.com)... 128.208.4.108
Connecting to pjreddie.com (pjreddie.com)[128.208.4.108]:443... connected.
HTTP request sent, awaiting response... 200 OK
length: 248007048 (237M) [application/octet-stream]
Saving to: 'yolov3.weights'

yolov3.weights  100%[=====] 236.52M  20.8MB/s  in 12s
2023-08-06 04:19:25 (19.4 MB/s) - 'yolov3.weights' saved [248007048/248007048]

2023-08-06 04:19:25-- https://raw.githubusercontent.com/pjreddie/darknet/master/cfg/yolov3.cfg
Resolving raw.githubusercontent.com (raw.githubusercontent.com)... 185.199.108.133, 185.199.109.133, 185.199.110.133, ...
Connecting to raw.githubusercontent.com (raw.githubusercontent.com)[185.199.108.133]:443... connected.
HTTP request sent, awaiting response... 200 OK
length: 8342 (8.1K) [text/plain]
Saving to: 'yolov3.cfg'

yolov3.cfg  100%[=====] 8.15K  --KB/s  in 0s
2023-08-06 04:19:25 (79.7 MB/s) - 'yolov3.cfg' saved [8342/8342]
```

```
!wget https://raw.githubusercontent.com/pjreddie/darknet/master/data/coco.names
```

Python

Downloads 'coco.names' file from the given URL using wget.

```
if not os.path.exists('Yolo_Results'):
    os.makedirs('Yolo_Results')
yolo_save_path = "/content/Yolo_Results"

classes = open(yolo_path+"coco.names").read().strip().split("\n")
weights = os.path.join(yolo_path+"yolov3.weights")
cfg = os.path.join(yolo_path+"yolov3.cfg")
np.random.seed(42)
yolo_colors = np.random.randint(0, 255, size=(len(classes), 3), dtype="uint8")
```

This code creates a directory 'Yolo_Results', sets up paths for YOLOv3 detection using COCO model, reads class names, loads pre-trained weights, and generates random colors for visualization.

```
net = cv2.dnn.readNetFromDarknet(cfg, weights)
detection_time = []
```

Python

In this code snippet, a YOLO neural network model is loaded using the OpenCV library. The 'cv2.dnn.readNetFromDarknet' function is used to read a YOLO model from its configuration file ('cfg') and corresponding pre-trained weights ('weights'). The 'detection_time' list is initialized to store the time taken for each detection operation. This code sets up the YOLO model for further object detection operations.

```
for im in images :
    img = cv2.imread(im,cv2.IMREAD_UNCHANGED)
    # print(img)
    (h, w) = img.shape[:2]
    output = net.getLayerNames()
    output = [output[i]-1] for i in net.getUnconnectedOutLayers()
    blob = cv2.dnn.blobFromImage(img, 1 / 255.0, (416, 416), swapRB=True, crop=False)
    net.setInput(blob)
    t = time.time()
    layerOutputs = net.forward(output)
    detection_time.append(time.time()-t)
    bbox_arr = []
    confidence_arr = []
    class_names_arr = []
    for lo in layerOutputs:
        for op in lo:
            marks = op[5:]
            class_name = np.argmax(marks)
            cfd = marks[class_name]
```

```

if cfd > confidence_threshold:
    box = op[0:4] * np.array([w, h, w, h])
    (X, Y, W, H) = box.astype("int")
    x = int(X - (W / 2))
    y = int(Y - (H / 2))
    bbox_arr.append([x, y, int(W), int(H)])
    confidence_arr.append(float(cfd))
    class_names_arr.append(class_name)
idx_arr = cv2.dnn.INDEXES(bbox_arr, confidence_arr, confidence_threshold, nms_threshold)
op_file = open(yolo_save_path+"/"+im.split('/')[:-1][-1]+".txt", "w")
if len(idx_arr) > 0:
    for i in idx_arr.flatten():
        (x, y) = (bbox_arr[i][0], bbox_arr[i][1])
        (w, h) = (bbox_arr[i][2], bbox_arr[i][3])
        clr = [int(c) for c in yolo_colors[class_names_arr[i]]]
        cv2.rectangle(img, (x, y), (x + w, y + h), clr, 2)
        txt = "{}: {:.4f}".format(class_names_arr[i], confidence_arr[i])
        cv2.putText(img, txt, (x + 5, cv2.dnn.INDEXES(bbox_arr, confidence_arr, confidence_threshold, nms_threshold) + 5), clr, 2)
        op_file.write("{} {} {} {} {} {} \n".format(class_names_arr[i], confidence_arr[i], x, y, x+w, y+h))
op_file.close()
cv2.imwrite(yolo_save_path+"/"+im.split('/')[:-1][-1], img)

```

This code iterates through a list of image file paths. For each image, it reads and processes the image using OpenCV's Deep Neural Network (DNN) module and a pre-trained YOLO model. Detected objects with confidence above a threshold are extracted, and non-maximum suppression is applied to remove redundant detections. Bounding boxes are drawn on the image and saved along with detection information in a text file. The processed image is saved to an output folder.

```

import os
import matplotlib.pyplot as plt
from matplotlib.image import imread

# Path to the folder containing the images
folder_path = 'Yolo_Results'

# List all files in the folder
image_files = [f for f in os.listdir(folder_path) if f.endswith('.jpg')]

# Check if there are any image files
if not image_files:
    print("No image files found in the folder.")
else:
    # Display the first image
    image_path = os.path.join(folder_path, image_files[0])
    image = imread(image_path)

    plt.imshow(image)
    plt.title("Image from 'Yolo_Results' Folder")
    plt.axis('off')
    plt.show()

```

The code reads image files from a specified folder using the 'os' library, displays the first image using 'matplotlib', and provides a title while removing the axis labels for visualization.



Launched an EC2 instance on AWS to host the model and data for inference. After SSH into the server, the main folders are:

- input** - Contains raw video files to run object detection on. For example, trafficcam.mp4.
- datadata** - Stores extracted frames from the input videos in JPEG format.
- cfgcfg** - Holds the YOLOv3 model configuration file.
- weightsweights** - Contains pre-trained weights for the YOLOv3 model.

sortresults- Output folder where object detection results on frames are saved.

The main command to run is:

```
python3 extract.py --video input/trafficcamm.mp4 --cfg cfg/yolov3 --weights yolov3.weights
```

This extracts all frames from trafficcamm.mp4 into the data folder as JPEGs. Then it runs YOLOv3 object detection on each frame using the cfg and weights files. The detections are saved as bounding box images in sortresults.

This pipeline allows new traffic camera videos to be continually processed on EC2 using the YOLOv3 model. The sortresults folder provides the output with object detections visualized on each frame. Storing the data and results on S3 could enable more scalable processing.

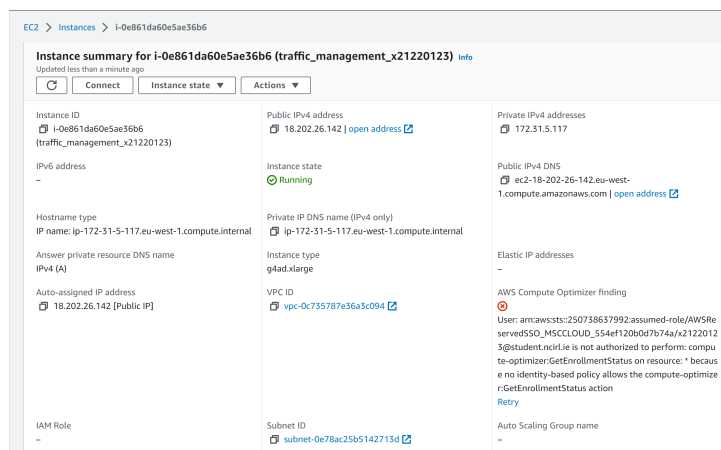
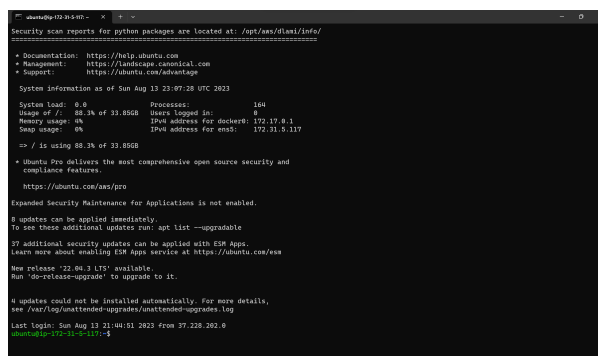


Figure 1: Ec2 instance



http://18.202.26.142:5000/traffic/playVideo/video_XXX.mp4
When accessed, this URL will download the MP4 video with object detections to the local machine.

