

AAI521_Team6_Final_V5

December 11, 2023

1 AAI521-Final Project: Team 6

- Christi Moncrief
- Adam Graves
- Reed Oken

##Edge IoT Device for Detecting and Counting People in a Region Of Interest

#####

```
[ ]: #@title 1: Install ultralytics for YOLO library
```

```
!pip install ultralytics
```

```
Requirement already satisfied: ultralytics in /usr/local/lib/python3.10/dist-packages (8.0.226)
```

```
Requirement already satisfied: matplotlib>=3.3.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from ultralytics) (3.7.1)
```

```
Requirement already satisfied: numpy>=1.22.2 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (1.23.5)
```

```
Requirement already satisfied: opencv-python>=4.6.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from ultralytics) (4.8.0.76)
```

```
Requirement already satisfied: pillow>=7.1.2 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (9.4.0)
```

```
Requirement already satisfied: pyyaml>=5.3.1 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (6.0.1)
```

```
Requirement already satisfied: requests>=2.23.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from ultralytics) (2.31.0)
```

```
Requirement already satisfied: scipy>=1.4.1 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (1.11.4)
```

```
Requirement already satisfied: torch>=1.8.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (2.1.0+cu118)
```

```
Requirement already satisfied: torchvision>=0.9.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from ultralytics) (0.16.0+cu118)
```

```
Requirement already satisfied: tqdm>=4.64.0 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (4.66.1)
```

```
Requirement already satisfied: pandas>=1.1.4 in /usr/local/lib/python3.10/dist-packages (from ultralytics) (1.5.3)
```

```
Requirement already satisfied: seaborn>=0.11.0 in
```

/usr/local/lib/python3.10/dist-packages (from ultralytics) (0.12.2)
 Requirement already satisfied: psutil in /usr/local/lib/python3.10/dist-packages
 (from ultralytics) (5.9.5)
 Requirement already satisfied: py-cpuinfo in /usr/local/lib/python3.10/dist-
 packages (from ultralytics) (9.0.0)
 Requirement already satisfied: thop>=0.1.1 in /usr/local/lib/python3.10/dist-
 packages (from ultralytics) (0.1.1.post2209072238)
 Requirement already satisfied: contourpy>=1.0.1 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (1.2.0)
 Requirement already satisfied: cycler>=0.10 in /usr/local/lib/python3.10/dist-
 packages (from matplotlib>=3.3.0->ultralytics) (0.12.1)
 Requirement already satisfied: fonttools>=4.22.0 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (4.45.1)
 Requirement already satisfied: kiwisolver>=1.0.1 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (1.4.5)
 Requirement already satisfied: packaging>=20.0 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (23.2)
 Requirement already satisfied: pyparsing>=2.3.1 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (3.1.1)
 Requirement already satisfied: python-dateutil>=2.7 in
 /usr/local/lib/python3.10/dist-packages (from matplotlib>=3.3.0->ultralytics)
 (2.8.2)
 Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-
 packages (from pandas>=1.1.4->ultralytics) (2023.3.post1)
 Requirement already satisfied: charset-normalizer<4,>=2 in
 /usr/local/lib/python3.10/dist-packages (from requests>=2.23.0->ultralytics)
 (3.3.2)
 Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-
 packages (from requests>=2.23.0->ultralytics) (3.6)
 Requirement already satisfied: urllib3<3,>=1.21.1 in
 /usr/local/lib/python3.10/dist-packages (from requests>=2.23.0->ultralytics)
 (2.0.7)
 Requirement already satisfied: certifi>=2017.4.17 in
 /usr/local/lib/python3.10/dist-packages (from requests>=2.23.0->ultralytics)
 (2023.11.17)
 Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-
 packages (from torch>=1.8.0->ultralytics) (3.13.1)
 Requirement already satisfied: typing-extensions in
 /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytics) (4.5.0)
 Requirement already satisfied: sympy in /usr/local/lib/python3.10/dist-packages
 (from torch>=1.8.0->ultralytics) (1.12)
 Requirement already satisfied: networkx in /usr/local/lib/python3.10/dist-
 packages (from torch>=1.8.0->ultralytics) (3.2.1)

Requirement already satisfied: jinja2 in /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytics) (3.1.2)
Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytics) (2023.6.0)
Requirement already satisfied: triton==2.1.0 in /usr/local/lib/python3.10/dist-packages (from torch>=1.8.0->ultralytics) (2.1.0)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.7->matplotlib>=3.3.0->ultralytics) (1.16.0)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from jinja2->torch>=1.8.0->ultralytics) (2.1.3)
Requirement already satisfied: mpmath>=0.19 in /usr/local/lib/python3.10/dist-packages (from sympy->torch>=1.8.0->ultralytics) (1.3.0)

```
[ ]: #@title 1.1: Loading related libraries
import cv2
from ultralytics import YOLO

import matplotlib.pyplot as plt

import pandas as pd
import numpy as np
import os
import subprocess
from tqdm.notebook import tqdm

from IPython.display import Video, display, HTML, Javascript, Image
import datetime

from base64 import b64encode, b64decode

from google.colab.output import eval_js
from google.colab.patches import cv2_imshow

from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
[ ]: #@title 2: Loading a YOLO model
model = YOLO('yolov8x.pt')

# Enhance the YOLO model with additional knowledge transfer to help with unique
↳ lighting or busy ROI
model.train(data="coco128.yaml", epochs=3) # train the model
```

```
#geting names from classes
dict_classes = model.model.names
```

Ultralytics YOLOv8.0.226 Python-3.10.12 torch-2.1.0+cu118 CUDA:0 (Tesla T4, 15102MiB)

```
engine/trainer: task=detect, mode=train, model=yolov8x.pt,
data=coco128.yaml, epochs=3, patience=50, batch=16, imgsz=640, save=True,
save_period=-1, cache=False, device=None, workers=8, project=None, name=train2,
exist_ok=False, pretrained=True, optimizer=auto, verbose=True, seed=0,
deterministic=True, single_cls=False, rect=False, cos_lr=False, close_mosaic=10,
resume=False, amp=True, fraction=1.0, profile=False, freeze=None,
overlap_mask=True, mask_ratio=4, dropout=0.0, val=True, split=val,
save_json=False, save_hybrid=False, conf=None, iou=0.7, max_det=300, half=False,
dnn=False, plots=True, source=None, vid_stride=1, stream_buffer=False,
visualize=False, augment=False, agnostic_nms=False, classes=None,
retina_masks=False, show=False, save_frames=False, save_txt=False,
save_conf=False, save_crop=False, show_labels=True, show_conf=True,
show_boxes=True, line_width=None, format=torchscript, keras=False,
optimize=False, int8=False, dynamic=False, simplify=False, opset=None,
workspace=4, nms=False, lr0=0.01, lrf=0.01, momentum=0.937, weight_decay=0.0005,
warmup_epochs=3.0, warmup_momentum=0.8, warmup_bias_lr=0.1, box=7.5, cls=0.5,
dfl=1.5, pose=12.0, kobj=1.0, label_smoothing=0.0, nbs=64, hsv_h=0.015,
hsv_s=0.7, hsv_v=0.4, degrees=0.0, translate=0.1, scale=0.5, shear=0.0,
perspective=0.0, flipud=0.0,fliplr=0.5, mosaic=1.0, mixup=0.0, copy_paste=0.0,
cfg=None, tracker=botsort.yaml, save_dir=runs/detect/train2
```

	from	n	params	module
arguments				
0	-1	1	2320	ultralytics.nn.modules.conv.Conv
[3, 80, 3, 2]				
1	-1	1	115520	ultralytics.nn.modules.conv.Conv
[80, 160, 3, 2]				
2	-1	3	436800	ultralytics.nn.modules.block.C2f
[160, 160, 3, True]				
3	-1	1	461440	ultralytics.nn.modules.conv.Conv
[160, 320, 3, 2]				
4	-1	6	3281920	ultralytics.nn.modules.block.C2f
[320, 320, 6, True]				
5	-1	1	1844480	ultralytics.nn.modules.conv.Conv
[320, 640, 3, 2]				
6	-1	6	13117440	ultralytics.nn.modules.block.C2f
[640, 640, 6, True]				
7	-1	1	3687680	ultralytics.nn.modules.conv.Conv
[640, 640, 3, 2]				
8	-1	3	6969600	ultralytics.nn.modules.block.C2f
[640, 640, 3, True]				
9	-1	1	1025920	ultralytics.nn.modules.block.SPPF

```

[640, 640, 5]
 10          -1  1          0 torch.nn.modules.upsampling.Upsample
[None, 2, 'nearest']
 11          [-1, 6]  1          0 ultralytics.nn.modules.conv.Concat
[1]
 12          -1  3   7379200 ultralytics.nn.modules.block.C2f
[1280, 640, 3]
 13          -1  1          0 torch.nn.modules.upsampling.Upsample
[None, 2, 'nearest']
 14          [-1, 4]  1          0 ultralytics.nn.modules.conv.Concat
[1]
 15          -1  3   1948800 ultralytics.nn.modules.block.C2f
[960, 320, 3]
 16          -1  1    922240 ultralytics.nn.modules.conv.Conv
[320, 320, 3, 2]
 17          [-1, 12]  1          0 ultralytics.nn.modules.conv.Concat
[1]
 18          -1  3   7174400 ultralytics.nn.modules.block.C2f
[960, 640, 3]
 19          -1  1   3687680 ultralytics.nn.modules.conv.Conv
[640, 640, 3, 2]
 20          [-1, 9]  1          0 ultralytics.nn.modules.conv.Concat
[1]
 21          -1  3   7379200 ultralytics.nn.modules.block.C2f
[1280, 640, 3]
 22          [15, 18, 21]  1   8795008 ultralytics.nn.modules.head.Detect
[80, [320, 640, 640]]
Model summary: 365 layers, 68229648 parameters, 68229632 gradients, 258.5 GFLOPs

```

Transferred 595/595 items from pretrained weights

TensorBoard: Start with 'tensorboard --logdir runs/detect/train2',
view at <http://localhost:6006/>

Freezing layer 'model.22.dfl.conv.weight'

AMP: running Automatic Mixed Precision (AMP) checks with YOLOv8n...

AMP: checks passed

train: Scanning /content/datasets/coco128/labels/train2017.cache...

126 images, 2 backgrounds, 0 corrupt: 100%| | 128/128 [00:00<?, ?it/s]

augmentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01,
blur_limit=(3, 7)), ToGray(p=0.01), CLAHE(p=0.01, clip_limit=(1, 4.0),
tile_grid_size=(8, 8))

val: Scanning /content/datasets/coco128/labels/train2017.cache...

126 images, 2 backgrounds, 0 corrupt: 100%| | 128/128 [00:00<?, ?it/s]

Plotting labels to runs/detect/train2/labels.jpg...

optimizer: 'optimizer=auto' found, ignoring 'lr0=0.01' and
'momentum=0.937' and determining best 'optimizer', 'lr0' and 'momentum'
automatically...

optimizer: AdamW(lr=0.000119, momentum=0.9) with parameter groups
 97 weight(decay=0.0), 104 weight(decay=0.0005), 103 bias(decay=0.0)
 Image sizes 640 train, 640 val
 Using 2 dataloader workers
 Logging results to runs/detect/train2
 Starting training for 3 epochs...

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
1/3	13.8G	0.8623	0.8033	1.156	228	640:
100%	8/8	[00:11<00:00, 1.44s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.12it/s]				
	all	128	929	0.816	0.733	0.833
0.669						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
2/3	14.3G	0.8665	0.6925	1.13	151	640:
100%	8/8	[00:10<00:00, 1.29s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:03<00:00, 1.11it/s]				
	all	128	929	0.877	0.746	0.855
0.69						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
3/3	14.3G	0.8381	0.6663	1.119	164	640:
100%	8/8	[00:10<00:00, 1.28s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.01s/it]				
	all	128	929	0.875	0.758	0.85
0.688						

3 epochs completed in 0.018 hours.

Optimizer stripped from runs/detect/train2/weights/last.pt, 136.9MB

Optimizer stripped from runs/detect/train2/weights/best.pt, 136.9MB

Validating runs/detect/train2/weights/best.pt...

Ultralytics YOLOv8.0.226 Python-3.10.12 torch-2.1.0+cu118 CUDA:0 (Tesla T4, 15102MiB)

Model summary (fused): 268 layers, 68200608 parameters, 0 gradients, 257.8 GFLOPs

	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%		4/4	[00:07<00:00,	1.77s/it]		
0.689	all	128	929	0.877	0.746	0.854
0.688	person	128	254	0.951	0.705	0.887
0.597	bicycle	128	6	0.776	0.667	0.739
0.374	car	128	46	1	0.39	0.673
0.812	motorcycle	128	5	0.987	1	0.995
0.933	airplane	128	6	0.953	1	0.995
0.768	bus	128	7	0.897	0.714	0.864
0.995	train	128	3	0.877	1	0.995
0.47	truck	128	12	0.784	0.417	0.682
0.573	boat	128	6	1	0.59	0.816
0.325	traffic light	128	14	1	0.337	0.544
0.995	stop sign	128	2	0.875	1	0.995
0.636	bench	128	9	1	0.629	0.852
0.695	bird	128	16	0.967	1	0.995
0.924	cat	128	4	0.906	1	0.995
0.912	dog	128	9	0.894	0.942	0.984
0.799	horse	128	2	0.854	1	0.995
0.851	elephant	128	17	0.967	0.941	0.953
0.995	bear	128	1	0.765	1	0.995
0.995	zebra	128	4	0.908	1	0.995
0.825	giraffe	128	9	0.948	1	0.995
	backpack	128	6	1	0.638	0.752

0.528	umbrella	128	18	0.954	0.833	0.943
0.704	handbag	128	19	0.838	0.546	0.699
0.5	tie	128	7	1	0.792	0.86
0.766	suitcase	128	4	0.933	1	0.995
0.648	frisbee	128	5	0.787	0.8	0.806
0.731	skis	128	1	0.809	1	0.995
0.796	snowboard	128	7	0.941	0.857	0.864
0.821	sports ball	128	6	0.633	0.584	0.535
0.361	kite	128	10	0.949	0.4	0.513
0.18	baseball bat	128	4	0.989	1	0.995
0.592	baseball glove	128	7	0.657	0.429	0.401
0.293	skateboard	128	5	0.614	0.6	0.762
0.548	tennis racket	128	7	1	0.699	0.721
0.412	bottle	128	18	0.668	0.447	0.706
0.485	wine glass	128	16	0.972	0.5	0.71
0.536	cup	128	36	0.928	0.717	0.924
0.681	fork	128	6	0.742	0.5	0.725
0.538	knife	128	16	0.915	0.675	0.896
0.618	spoon	128	22	0.885	0.636	0.753
0.633	bowl	128	28	0.874	0.786	0.831
0.732	banana	128	1	1	0	0.995
0.995	sandwich	128	2	0.808	1	0.995
0.995	orange	128	4	0.611	1	0.912
0.717	broccoli	128	11	1	0.366	0.592

0.451						
	carrot	128	24	0.826	0.792	0.875
0.671						
	hot dog	128	2	0.851	1	0.995
0.995						
	pizza	128	5	0.882	1	0.995
0.882						
	donut	128	14	0.776	1	0.995
0.933						
	cake	128	4	0.86	1	0.995
0.905						
	chair	128	35	0.715	0.743	0.828
0.596						
	couch	128	6	0.991	1	0.995
0.853						
	potted plant	128	14	0.892	0.591	0.889
0.717						
	bed	128	3	0.92	1	0.995
0.805						
	dining table	128	13	1	0.715	0.872
0.796						
	toilet	128	2	0.912	1	0.995
0.739						
	tv	128	2	0.817	1	0.995
0.895						
	laptop	128	3	0.616	0.667	0.863
0.792						
	mouse	128	2	1	0	0.663
0.333						
	remote	128	8	1	0.66	0.762
0.659						
	cell phone	128	8	0.969	0.625	0.697
0.436						
	microwave	128	3	0.917	1	0.995
0.93						
	oven	128	5	0.537	0.4	0.298
0.233						
	sink	128	6	0.667	0.5	0.643
0.514						
	refrigerator	128	5	0.878	1	0.995
0.845						
	book	128	29	0.772	0.414	0.562
0.35						
	clock	128	9	0.956	0.889	0.975
0.831						
	vase	128	2	0.797	1	0.995
0.995						
	scissors	128	1	1	0	0.995

```

0.199
teddy bear      128      21      0.964      0.857      0.977
0.72
toothbrush      128       5      0.921       1      0.995
0.868
Speed: 1.1ms preprocess, 23.2ms inference, 0.0ms loss, 2.7ms postprocess per
image
Results saved to runs/detect/train2

```

```

[ ]: #@title 2.1: Load video (emulate capture)
# Load video (emulate capture)
video_name = 'City_Hall-IOT1'
encoding_type = '.mp4'
video_folder = '/content/drive/MyDrive/AAI-521/Final/Test/'

video_path = video_folder + video_name + encoding_type
video = cv2.VideoCapture(video_path)

[ ]: #@title 3: Define required functions for video processing
# process frames
def resize_frame(frame, scale_percent):
    """Function to resize frame"""
    # resize image
    width = int(frame.shape[1] * scale_percent / 100)
    height = int(frame.shape[0] * scale_percent / 100)
    dim = (width, height)
    # resize image
    resized = cv2.resize(frame, dim, interpolation = cv2.INTER_AREA)
    return resized

# Filter the history of tracked objects
def filter_tracks(centers, patience):
    """Function to filter the history of tracked objects"""
    filter_dict = {}
    for k, i in centers.items():
        d_frames = i.items()
        filter_dict[k] = dict(list(d_frames)[-patience:])

    return filter_dict

# Update tracked objects
def update_tracking(centers_old, obj_center, thr_centers, lastKey, frame,
    frame_max):
    is_new = 0 # Reset count
    lastpos = [(k, list(center.keys())[-1], list(center.values())[-1]) for k,
    center in centers_old.items()]

```

```

    lastpos = [(i[0], i[2]) for i in lastpos if abs(i[1] - frame) <= frame_max]
    ↪ # Check position within each frame
    # Calculating distance from existing centers points - calculate if same or
    ↪ new person
    previous_pos = [(k,obj_center) for k,centers in lastpos if (np.linalg.
    ↪ norm(np.array(centers) - np.array(obj_center)) < thr_centers)]
    # if distance less than a threshold, it will update its positions
    if previous_pos:
        id_obj = previous_pos[0][0]
        centers_old[id_obj][frame] = obj_center

    # Else a new ID will be set to the given object
    else:
        if lastKey:
            last = lastKey.split('D')[1]
            id_obj = 'ID' + str(int(last)+1)
        else:
            id_obj = 'ID0'

        is_new = 1
        centers_old[id_obj] = {frame:obj_center}
        lastKey = list(centers_old.keys())[-1]

    return centers_old, id_obj, is_new, lastKey

def plt_pretty_image(image, label=None):
    """Function to display an image using matplotlib.pyplot.imshow with plt
    ↪ graphing features (Grid and x/y ticks removed)"""
    plt.xticks([])
    plt.yticks([])
    plt.grid(False)
    plt.imshow(image, cmap='gray')
    plt.xlabel(label)

def video_to_base64(file_path):
    with open(file_path, 'rb') as video_file:
        encoded_string = b64encode(video_file.read()).decode()
    return encoded_string

```

[]: #@title 3.1: Configurations for ROI objects

```

# Scaling percentage of original frame
scale_percent = 100
# model confidence level
conf_level = 0.6
# Threshold of centers ( old\new)
thr_centers = 30 # Pixels between objects (modify per test MP4)

```

```

#Number of max frames to consider a object lost
frame_max = 5
# Number of max tracked centers stored
patience = 100
# ROI area color transparency
alpha = 0.2

```

```

[ ]: #@title 4.1: Set person class to detect in YOLO
# Objects to detect Yolo
class_IDS = [0] #Persons

# Auxiliary variables
centers_old = {}
obj_id = 0
end = []
frames_list = []
count_p = 0
lastKey = ''

```

```

[ ]: #@title 4.2: Settings for input of video
# Original informations of video
height = int(video.get(cv2.CAP_PROP_FRAME_HEIGHT))
width = int(video.get(cv2.CAP_PROP_FRAME_WIDTH))
fps = video.get(cv2.CAP_PROP_FPS)
frame_count = int(video.get(cv2.CAP_PROP_FRAME_COUNT))
print(f'Original dimensions: {(width, height)}')
print(f'Original FPS: {fps}')

# Scaling Video for better performance
if scale_percent != 100:
    print('Scaling change may cause errors in pixels lines ')
    width = int(width * scale_percent / 100)
    height = int(height * scale_percent / 100)
    print(f'Dimension scaled: {(width, height)}')

```

Original dimensions: (1920, 1080)
Original FPS: 25.0

```

[ ]: #@title 4.3: Settings on video output
# Settings for video output
output_video_path = video_name + '_result.mp4'
VIDEO_CODEC = "MP4V" #set to MP4 codec
tmp_output_path = 'tmp_' + output_video_path

fourcc = cv2.VideoWriter_fourcc(*VIDEO_CODEC)
output_video = cv2.VideoWriter(
    tmp_output_path,

```

```

        fourcc=fourcc,
        fps=fps,
        frameSize=(width, height)
    )

```

```

[ ]: #@title 4.4 ROI area setup
custom_roi = False

# default full frame
if not custom_roi:
    x_min = 0
    x_max = x_min + width
    y_min = 0
    y_max = y_min + height
else:
    x_min = 100
    x_max = 500
    y_min = 50
    y_max = 300

blur_factor = 0

# font scaling
font_size = min(width, height) // 300
font_thickness = max(6, font_size // 2)

```

```

[ ]: #@title 5: Executing Recognition in an ROI section

# Initialize DataFrame to store results
results_df = pd.DataFrame(columns=['Frame', 'Timestamp', 'People_Count', 'Confidence_Scores'])

for i in tqdm(range(frame_count)):

    # reading frame from video
    ret, frame = video.read()

    if not ret:
        continue

    # Applying resizing of read frame
    if scale_percent != 100:
        frame = resize_frame(frame, scale_percent)

    # Apply Gaussian Blur to the frame
    frame = cv2.cvtColor(frame, cv2.COLOR_BGR2RGB)
    frame = cv2.GaussianBlur(frame, (3, 3), blur_factor)

```

```

# Setup ROI
area_roi = [np.
↪array([(x_min,y_min),(x_min,y_max),(x_max,y_max),(x_max,y_min)], np.int32)]
roi_frame = frame[y_min:y_max, x_min:x_max]

# Getting predictions
y_hat = model.predict(roi_frame, conf = conf_level, classes = class_IDS,
↪device = 0, verbose = False)

# Update total count of people
count_p = 0
confidence_scores = [] #Set for Excel output

# Fetch current time
current_time = datetime.datetime.now()
time_str = current_time.strftime("%H:%M:%S") # Format time as HH:MM:SS

text_size = cv2.getTextSize(time_str, cv2.FONT_HERSHEY_SIMPLEX, font_size/
↪2, font_thickness//2)[0]
text_x = int(frame.shape[1] - text_size[0] - (frame.shape[1] * 0.01))
text_y = int(frame.shape[0] * 0.1)

cv2.putText(frame, time_str, (text_x, text_y),
            cv2.FONT_HERSHEY_SIMPLEX, font_size/2, (255, 255, 255),
↪font_thickness//2)

# Getting the bounding boxes, confidence and classes of the recognize
↪objects in the current frame.
boxes_data = y_hat[0].boxes.xyxy.cpu().numpy() #Move into CPU memory
confidences = y_hat[0].boxes.conf.cpu().numpy()
classes = y_hat[0].boxes.cls.cpu().numpy()

# Storing the above information in a dataframe, setting the min and max
↪values
positions_frame = pd.DataFrame(boxes_data, columns=['xmin', 'ymin', 'xmax',
↪'ymax'])
positions_frame['conf'] = confidences
positions_frame['class'] = classes

#Translating the numeric class labels to text
labels = [dict_classes[i] for i in classes]

#handle tracking logic
for ix, row in enumerate(positions_frame.iterrows()):
    xmin, ymin, xmax, ymax, confidence, category = row[1].astype('int')

```

```

count_p += 1

confidence_scores.append(confidence) #Append data for Excel file
# Calculating the center of the bounding box
center_x, center_y = int(((xmax + xmin) / 2)), int(((ymax + ymin) / 2))
# Drawing bounding box for every detection
cv2.rectangle(roi_frame, (xmin, ymin), (xmax, ymax), (0, 0, 255), 2)

# Updating the tracking for each object
centers_old, id_obj, is_new, lastKey = update_tracking(centers_old,
↳(center_x, center_y), thr_centers, lastKey, i, frame_max)

# Drawing additional tracking info (like circles) if needed (New)
for center_x, center_y in centers_old[id_obj].values():
    cv2.circle(roi_frame, (center_x, center_y), 5, (0, 255, 0), -1) # Using
↳a different color for clarity

#Updating people in ROI
count_p+=is_new

#Drawing above the bounding-box the name of class recognized.
cv2.putText(img=roi_frame,
            text=id_obj + ':' + str(np.round(confidences[ix], 2)),
            org=(xmin, ymin - 40), # Adjust this value to position the text
↳higher
            fontFace=cv2.FONT_HERSHEY_TRIPLEX,
            fontScale=0.9,
            color=(0, 0, 255),
            thickness=1)

count_str = f'No. of People in ROI: {count_p}'
baseline = cv2.getTextSize(count_str, cv2.FONT_HERSHEY_SIMPLEX, font_size,
↳font_thickness)[1]
text_x = int(frame.shape[1] * 0.05)
text_y = int(frame.shape[0] * 0.1) + baseline

cv2.putText(img=frame, text=count_str,
            org=(text_x, text_y), fontFace=cv2.FONT_HERSHEY_SIMPLEX,
            fontScale=font_size, color=(255, 255, 255),
↳thickness=font_thickness)

# Append results for this frame to DataFrame (Excel)
results_df = results_df.append({'Frame': i, 'Timestamp': time_str,
↳'People_Count': count_p, 'ID': lastKey}, ignore_index=True)

```

```

# Filtering tracks history
centers_old = filter_tracks(centers_old, patience)

#Drawing the ROI area
overlay = frame.copy() # make a copy of the original frame

cv2.polylines(overlay, pts = area_roi, isClosed = True, color=(255, 0, 0),thickness=2)
cv2.fillPoly(overlay, pts = area_roi, color = (255,255,0))
frame = cv2.addWeighted(overlay, alpha,frame , 1 - alpha, 0)

#Saving frames in a list
frames_list.append(frame)
#saving transformed frames in a output video formaat
output_video.write(frame)

# After processing all frames, export the DataFrame to Excel
results_df.to_excel('/content/people_count.xlsx', index=False)

#Releasing the video
output_video.release()

```

0%| | 0/341 [00:00<?, ?it/s]

```

[ ]: #@title 5.1: Process Video and Audio
# Check for Existing Output File and Remove if Present
if os.path.exists(output_video_path):
    os.remove(output_video_path)
# Process video and audio file
subprocess.run(
    ["ffmpeg", "-i",
    tmp_output_path,"-crf","18","-preset","veryfast","-hide_banner","-loglevel","error","-vcode
os.remove(tmp_output_path)

```

```

[ ]: #@title 5.2: Display of people and results _ Testing
frames_to_display = 24

for i in range(frames_to_display):
    frame = frames_list[i*10].copy()

    plt.figure(figsize=(7, 5))
    plt_pretty_image(frame, f'frame {i*20}')
    plt.show()

```




frame 0



frame 20



frame 40



frame 60



frame 80



frame 100



frame 120



frame 140



frame 160



frame 180



frame 200



frame 220



frame 240



frame 260



frame 280



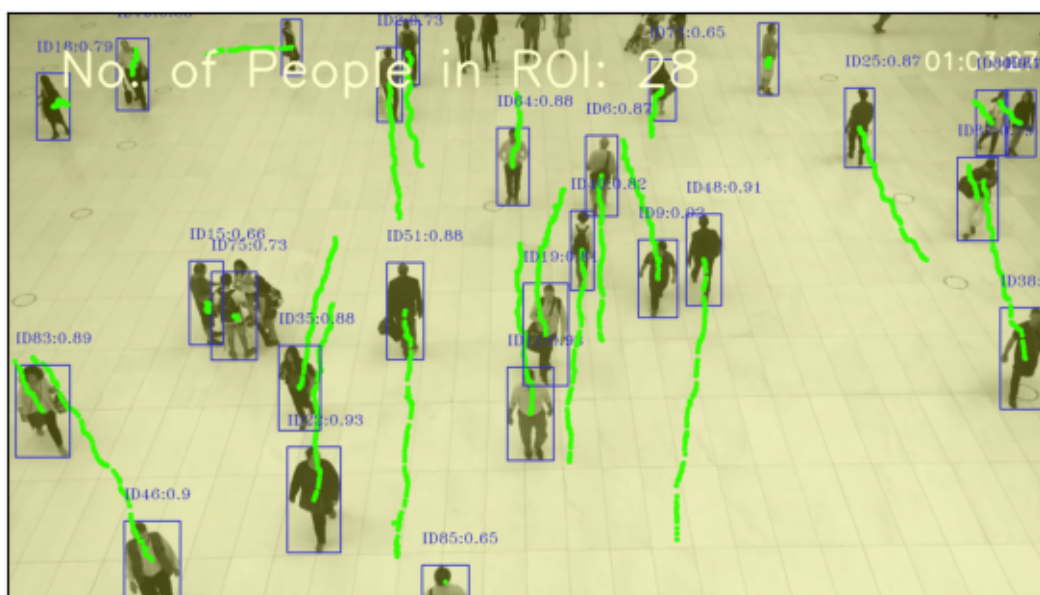
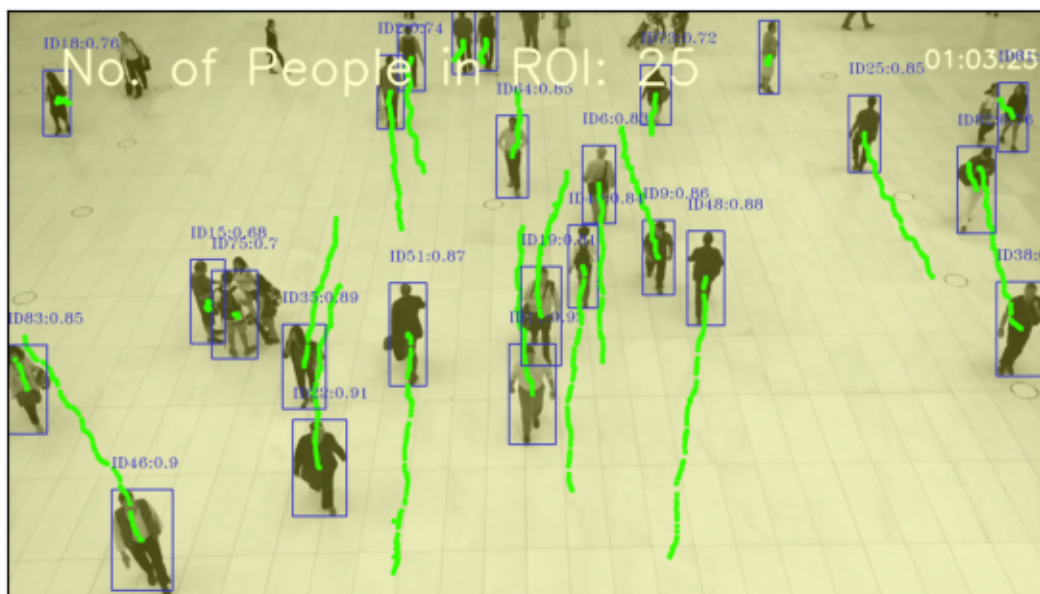
frame 300



frame 320



frame 340

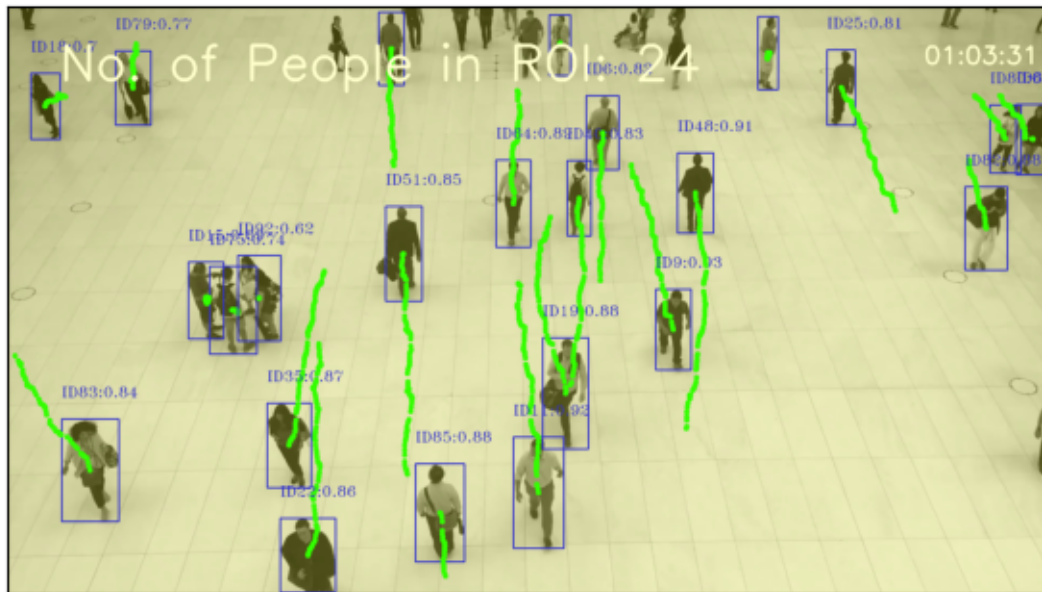




frame 400



frame 420



frame 440



frame 460

```
[ ]: #@title 5.3: Monitoring: Play the processed MP4
# preview output video
video_tag = f"""
<video width={width/2} height={height/2} controls>
```

```

        <source src="data:video/mp4;base64,{video_to_base64(output_video_path)}"
        ↪type="video/mp4">
    </video>
    """

```

```
[ ]: HTML(video_tag)
```

```
[ ]: <IPython.core.display.HTML object>
```

```

[ ]: #@title 6: Score the test set
# Load the test results file 'count_data.csv'
df = pd.read_excel('/content/drive/MyDrive/AAI-521/Final/Test/
    ↪Score-ManualCount-t6.xlsx')

# Initialize counters
TP = 0
FP = 0
FN = 0

# Calculate TP, FP, and FN
for index, row in df.iterrows():
    system_count = row['System Count']
    manual_count = row['M. Counted']

    TP += min(system_count, manual_count)
    FP += max(0, system_count - manual_count)
    FN += max(0, manual_count - system_count)

# Calculate Precision, Recall, and F1 Score
precision = TP / (TP + FP) if TP + FP > 0 else 0
recall = TP / (TP + FN) if TP + FN > 0 else 0
f1_score = 2 * (precision * recall) / (precision + recall) if precision +
    ↪recall > 0 else 0
#Print out the values
print("SCORE VALUES")
print(f"Precision: {precision:.4f}")
print(f"Recall: {recall:.4f}")
print(f"F1 Score: {f1_score:.4f}")

```

SCORE VALUES

Precision: 0.9948

Recall: 0.7265

F1 Score: 0.8397

#This section is for emulating a real-time Camera sensor in the IoT


```
[ ]: #@title 7: Installing required packages and testing the model
# import dependencies
from IPython.display import display, Javascript, Image
from google.colab.output import eval_js
from google.colab.patches import cv2_imshow
from base64 import b64decode, b64encode
import cv2
import numpy as np
import PIL
import io
import html
import time
import matplotlib.pyplot as plt
%matplotlib inline
```

```
[ ]: #@title 7.1: Open real-time camera capture with YOLO Object Detection
# Load the YOLOv8 model
model = YOLO("yolov8n.pt") # Adjust the model path as necessary
dict_classes = model.model.names # Class names

# Initialize camera stream
cap = cv2.VideoCapture(0) # Use appropriate index or URL for your camera

try:
    while True:
        # Read frame from camera
        ret, frame = cap.read()
        if not ret:
            print("Error: failed to capture image")
            break

        # Perform detection
        results = model.predict(frame)

        # Draw bounding boxes and labels
        for detection in results.xyxy[0]:
            # Extract data
            xmin, ymin, xmax, ymax, conf, cls = detection
            label = dict_classes[int(cls)]

            # Draw bounding box
            cv2.rectangle(frame, (int(xmin), int(ymin)), (int(xmax),
↪int(ymax)), (0, 255, 0), 2)

            # Draw label
            cv2.putText(frame, f"{label} {conf:.2f}", (int(xmin), int(ymin)-10),
                        cv2.FONT_HERSHEY_SIMPLEX, 0.5, (0, 255, 0), 2)
```

```
# Display the frame
cv2.imshow("Camera Feed", frame)

# Break loop on specific key press, e.g., 'q'
if cv2.waitKey(1) & 0xFF == ord('q'):
    break
finally:
    cap.release()
    cv2.destroyAllWindows()
```