```
In [1]:
         import cv2
         import torch
         import numpy as np
         from collections import OrderedDict
         from google.colab.patches import cv2 imshow
         import dlib
         import os
         from PIL import Image
         import matplotlib.pyplot as plt
         from pathlib import Path
In [ ]:
         # Loading dataset
         !wget http://www.cs.toronto.edu/~fidler/teaching/2018/CSC420 assign/project1.zip
         !unzip /content/project1.zip
In [ ]:
         !git clone https://github.com/ultralytics/yolov5.git
         !pip install -r yolov5/requirements.txt
         ## Download this for 50k images 8 epoches YOLOv5 medium model
         # !wget https://www.dropbox.com/s/imogdf12v59k9ov/yolo face 50k 8epochs.zip?dl=1
         ## Download this for 70k images 10 epoches YOLOv5 medium model
         !wget https://www.dropbox.com/s/8atygeezg7rvyf5/best.pt?dl=1
         ## Use this for 70k images 10 epoches YOLOv5 medium model
         !mv best.pt?dl=1 best.pt
         ## Use this for 70k images 10 epoches YOLOv5 medium model's weight
         ! python yolov5/detect.py --save-txt --weights /content/best.pt --source /content/proje
         !zip -r detect.zip /content/yolov5/runs/detect/
         ! zip -r clip_3_res.zip /content/yolov5/runs/detect/exp3
In [4]:
         # Below code is to convert Yolo's face detection result' txt to bbox
         # bbox is a dictionary with key being path to each frame and
         # value is each frames' corresponsing bounding boxes
         path = '/content/yolov5/runs/detect/exp/labels' # <--</pre>
                                           '/content/yolov5/runs/detect/expxxx/labels'
         # where xxx is the path where the detected labels are stored for yolo above.
         # For example, exp5 will be /content/yolov5/runs/detect/exp5/labels
         image path = '/content/project1/clip 3' # <-- This is where clip frames</pre>
                                                    are stored. Change to clip 1/2/3
         dir_list = os.listdir(path)
         bbox = \{\}
         for filename in dir list:
           f = open(os.path.join(path,filename),'r')
           lines = f.read().split('\n')
           f.close()
           img name = filename.split('.')[0]
           img = cv2.imread(os.path.join(image path,img name+'.jpg'))
           if img is not None:
             boxes = []
             for line in lines:
               f = filter(None, line.split(' '))
               part = list(f)
```

```
if len(part) > 0:
    x = float(part[1])
    y = float(part[2])
    w = float(part[3])
    h = float(part[4])
    X, Y = img.shape[1], img.shape[0]
    W = w * X
    H = h * Y
    xcenter = x * X
    ycenter = y * Y
    xmin = int(xcenter - W/2)
    ymin = int(ycenter - H/2)
    xmax = int(xcenter + W/2)
    ymax = int(ycenter + H/2)
    boxes.append((xmin, ymin, xmax, ymax))
bbox[os.path.join(image_path,img_name+'.jpg')] = boxes
```

### **Face Tracking Across Frames**

# Method 1: Accurate Scale Estimation for Robust Visual Tracking

Here I used the method introduced in the paper 'Accurate scale estimation for robust visual tracking.' Proceedings of the British Machine Vision Conference BMVC. 2014. (cite). This paper uses Correlation Filter Based Tracking(https://dl.acm.org/doi/book/10.5555/2520035) with Scale Estimation so that the algorithm can track an object and draw a bounding box that will adjust the size based on the object's size in the frame. Correlation Filter Based Tracking is similar to the template matching we told in the lecture, we input a frame with a bounding box, and the algorithm calculates a correlation filter based on the bounding box and then convolves the correlation filter in the next frame to find the pixel with a peak. The paper uses an additional algorithm to compute correlation filters with different sizes, so the algorithm is able to handle large-scale variations.

We implemented it using dlib.correlation\_tracker. This function is an implementation of the algorithm above.

We designed our algorithm as follows:

Loop over all the frames in order, and keep a list of dlib.correlation\_tracker for each bounding box. If there is no tracker, then we run the face detection algorithm and obtain the bounding box for that frame. After getting bounding boxes for that frame, for each bounding box, we initialize correlation\_tracker by inputting the frame and the bounding box. Then we keep tracking for all the following frames and evaluate the Peak To Sidelobe Ratio. This ratio is a metric that measures the peak sharpness of the correlation plane. If the ratio is lower than a threshold, then we abandon that tracker and assign all the bounding boxes that the tracker assigned for each frame belongs to same person (I assign an unique id for each person associated with the bounding box). Until we abandon all the trackers, we run the face detection algorithm and do the things above again.

```
In [5]: clip_img_paths = []
```

```
In [6]:
         img paths = clip img paths
         # Count how many face detection is runned
         count face detection = 0
         # Ordered dict to save a list of detected faces each time
         # {id: (tracker, [(frame_id, bounding_box), ...])}
         all faces = {}
         # finished tracking faces to store all the finished traked person ID's
         # bounding boxes
         finished tracking faces = {}
         next_face_id = 0
         # Hyperparameter
         face detection prob boundary = 0.8
         PSR boundary = 7
         frames_with_face_detection = set(bbox.keys())
         import time
         # Record time
         start = time.time()
         # Record the number of people that are currently tracking
         # If become zero then we need to run face detection again
         num of tracking = 0
         for i in range(len(img paths)):
           curr_path = str(img_paths[i])
           if curr_path in frames_with_face_detection:
             curr img = cv2.imread(str(curr path))
             # If we are not tracking any faces
             if num_of_tracking == 0:
               # Run face detection to get face bounding boxes
               boxes = bbox[str(curr_path)]
               count face detection += 1
               num of boxes = len(boxes)
               for j in range(num of boxes):
                 # initialize tracking process
                 curr box = boxes[j]
                 left_up_pt = curr_box[0], curr_box[1]
                 righ_dow_pt = curr_box[2], curr_box[3]
                 new_tracker = dlib.correlation_tracker()
                 new_tracker.start_track(curr_img,dlib.rectangle(left_up_pt[0],
                                left up pt[1],
                               righ dow pt[0],
                                righ dow pt[1]))
                 # Add new tracking face to face dictionary
                 all faces[next face id] = (new tracker, [(curr path, left up pt, \
                                                            righ dow pt)])
                 next face id += 1
                 num_of_tracking += 1
               key_list = all_faces.keys()
```

```
keys_to_pop = []
               for key in key list:
                 tracker = all faces[key][0]
                 trackingQuality = tracker.update(curr_img)
                 # http://dlib.net/python/index.html#dlib.correlation_tracker
                 The Peak to Sidelobe Ratio (PSR) is a metric that measures
                 the peak sharpness of the correlation plane. For the
                 estimation of the PSR the peak is located first. Then the mean
                 and standard deviation of the 20X20 sidelobe region -
                 excluding a 5X5 central mask centred at the peak are
                 computed. PSR is then calculated as follows:
                 (compues how unique the centre is)
                 if trackingQuality >= PSR_boundary:
                   tracked position = tracker.get position()
                   t x = int(tracked position.left())
                   t_y = int(tracked_position.top())
                   t w = int(tracked position.width())
                   t h = int(tracked position.height())
                   left_up_pt = (t_x, t_y)
                   righ_dow_pt = (t_x + t_w , t_y + t_h)
                   all faces[key][1].append((curr_path, left_up_pt, righ_dow_pt))
                 else:# the face is not being tracked, move to finished
                     #tracking ordered dic
                   finished_tracking_faces[key] = all_faces[key][1]
                   keys_to_pop.append(key)
                   num_of_tracking -= 1
               for key in keys to pop:
                 all faces.pop(key)
         # Move all face tracking from all_faces to finished_tracking_faces to
         # move all the faces that tracked to last frame
         for key in all faces.keys():
           finished_tracking_faces[key] = all_faces[key][1]
         # Record time
         end = time.time()
         print("time takes: ", end - start)
         print("face detection times: ", count_face_detection)
        time takes: 4.3610804080963135
        face detection times: 7
In [7]:
         # Draw bounding boxes based on the tracks and save them to save frames
         # save frames format: {"frame paths":open cv image}
         save frames = {}
         for key in finished tracking faces.keys():
           for box in finished_tracking_faces[key]:
             frame path = box[0]
             # Read image from path only if we have not draw bbox on it
             if frame_path in save_frames.keys():
               curr img = save frames[frame path]
             else:
               curr img = cv2.imread(str(frame path))
             left up pt = box[1]
             righ dow pt = box[2]
             # Draw bbox
```

In [8]:

```
cv2_imshow(save_frames[list(save_frames.keys())[22]])
```



```
In [9]:
    outpath = './content/result_clip_3'
    if not os.path.isdir(outpath):
        os.makedirs(outpath)
    for frame in save_frames:
            fname = os.path.split(frame)[-1]
            cv2.imwrite(os.path.join(outpath, fname),save_frames[frame])

In []:
    !ffmpeg -framerate 10 -pattern_type glob -i ./content/result_clip_3/'*.jpg' \
        -vf "pad=ceil(iw/2)*2:ceil(ih/2)*2"\
```

## Method 2 Compare the similarity between two

-c:v libx264 -pix\_fmt yuv420p out3.mp4

### frames bounding boxes and/or their content

Inspired by the shot detection algorithm, we can compare all the bounding boxes within a frame to all the bounding boxes in the following frame, and if they are similar or in a similar position, then we can say those two bounding boxes bound the same person. I tried to use two ways of calculating the similarity of the two bounding boxes:

- 1. Centroid of the bounding box.
- 2. Histogram differences between two bounding boxes' content.

#### Algorithm:

Loop over all the frames in increasing order. Inside the loop, run the face detection algorithm and obtain bounding boxes. Loop over all those bounding boxes and compare them to previous frames bounding boxes. If no previous frame existed or previous frames have no bounding box, then assign a unique person ID to each bounding box. If the previous frame existed, then compute the distance between this frame's bounding boxes and the previous frame's bounding boxes.

- 1. If two frames contain the same amount of bounding boxes or the current frame contains fewer bounding boxes, then assign the existing bounding box to the bounding box's ID in the previous frame, which two bounding boxes have the smallest distance.
- 2. If the current frame has more bounding boxes after all the previous frame's bounding boxes' id has been assigned. We assign a new person ID for all the rest of the current frame's bounding boxes.

```
In [11]:
          Score functions to compare two bounding boxes and/or its content
          def histogram differences(curr, prev, curr bbox, prev bbox, color mode="rgb"):
              The function will return histogram differences
              # Crop each image
              img1 = curr[int(curr_bbox[1]):int(curr_bbox[3]), int(curr_bbox[0]):int(curr_bbox[2])
              img2 = prev[int(prev_bbox[1]):int(prev_bbox[3]), int(prev_bbox[0]):int(prev_bbox[2])
              # Calculate histogram difference
              if color mode=="rgb":
                bgr split img1 = cv2.split(img1)
                bgr_split_img2 = cv2.split(img2)
                sum_hist_diff = 0
                for i in range(3):
                    hist_img1 = cv2.calcHist([bgr_split_img1[i]], [0], None, [256], [0, 256])
                    hist_img2 = cv2.calcHist([bgr_split_img2[i]], [0], None, [256], [0, 256])
                    cv2.normalize(hist_img1, hist_img1, norm_type=cv2.NORM_MINMAX)
                    cv2.normalize(hist_img2, hist_img2, norm_type=cv2.NORM_MINMAX)
                    hist diff = np.sum(np.abs(np.array(hist img1) - np.array(hist img2)))
                    sum hist diff = sum hist diff + hist diff
                return sum hist diff/3.0
                raise ValueError('Invalid color_mode')
          def centroid dist(curr, prev, curr bbox, prev bbox):
```

```
Compute the distance between two bounding box's centroid curr: current frame image prev: current frame image curr_bbox: one bounding box of the current frame prev_bbox: one bounding box of the current frame prev_bbox: one bounding box of the current frame """

curr_bbox_centroid = (curr_bbox[2]+curr_bbox[0])/2, (curr_bbox[3]+curr_bbox[1])/2 prev_bbox_centroid = (prev_bbox[2]+prev_bbox[0])/2, (prev_bbox[3]+prev_bbox[1])/2 dist = np.linalg.norm(np.array(curr_bbox_centroid)-np.array(prev_bbox_centroid)) return dist
```

```
In [12]:
          def compute distance(dist function, curr, prev, bboxs curr, bboxs prev, \
                                boundary=50):
            .....
            Using dist function compute the distance between two set of bounding boxes
            bboxs_curr, bboxs_prev,
            between two frames
            curr, prev,
            0.00
            # [(bbox1 index, [dist1, dist2...]), ()]
            bbox_dist = []
            bboxs_curr_assigned = []
            # Compute distance
            for bbox1_index in range(len(bboxs_curr)):
              dists = []
              for bbox2_index in range(len(bboxs_prev)):
                dist = dist function(curr, prev, bboxs curr[bbox1 index], \
                                      bboxs prev[bbox2 index])
                dists.append(dist)
              if dists == []:
                dists = [float('inf')]
              bbox dist.append((bbox1 index, dists))
            bbox_dist.sort(key=lambda s: min(s[1]))
            # Assign index to each current frame's bounding boxes
            # If No prev box id to be assigned then assign None
            assigned indexes = []
            for item in bbox dist:
              bbox1_index = item[0]
              dists = item[1]
              # To prevent assigned a previously occurred value
              for i in assigned indexes:
                dists[i] = float('inf')
              min_dist_bbox2_index = np.argmin(dists)
              assigned indexes.append(min dist bbox2 index)
              # If assigned then set the distance to infinity
              if dists[min dist bbox2 index] == float('inf'):
                min_dist_bbox2_index = None
              # If distance exceed boundary
              elif dists[min dist bbox2 index] >= boundary:
                min dist bbox2 index = None
              bboxs curr assigned.append((bbox1 index,min dist bbox2 index))
            return bboxs_curr_assigned
```

```
In [13]: img_paths = clip_img_paths
```

```
face detection result = bbox
# Ordered dict to save a list of detected faces each time
# {id: (tracker, [(frame_id, bounding_box), ...])}
dist function = centroid dist
# Difference boundary for centroid dist and histogram differences
boundary = 50
all_faces = {}
next face id = 0
# All the frames with face bounding boxes
frames with face detection = set(face detection result.keys())
# Record time
start = time.time()
# Record the number of people that are currently tracking
# If become zero then we need to run face detection again
num of tracking = 0
prev path = None
prev_img = None
bboxs_prev = []
# Loop over all frames
for i in range(len(img paths)):
  curr path = str(img paths[i])
  if curr_path in frames_with_face_detection:
    curr_img = cv2.imread(str(curr_path))
    # Run face detection
    boxes = face detection result[str(curr path)]
    num of boxes = len(boxes) # If no face detected
    assigned_indexies = []
    if num of boxes > 0:
      # Compute assignment when current frame has boxes
      assignment = compute_distance(dist_function,curr_img, prev_img, boxes, \
                                    bboxs prev)
      all faces[curr path] = {}
      curr bboxes indexes = set(range(len(boxes)))
      for item in assignment:
        bbox1 index = item[0]
        assigned indexies.append(bbox1 index)
        curr box = boxes[bbox1 index]
        left up pt = curr box[0], curr box[1]
        righ_dow_pt = curr_box[2], curr_box[3]
        min dist bbox2 index = item[1]
        # Case 2 if prev has more boxes
        # Case 3 if they are equal
        # Assign/ingerit previous box's ID
        if min dist bbox2 index is not None:
          prev box = bboxs prev[min dist bbox2 index]
          prev_left_up_pt = prev_box[0], prev_box[1]
          prev_righ_dow_pt = prev_box[2], prev_box[3]
          prev_box = prev_left_up_pt, prev_righ_dow_pt
          prev assigned id = all faces[prev path][prev box]
          box pt = left up pt, righ dow pt
          all_faces[curr_path][box_pt] = prev_assigned_id
        # Case 1 if curr has more boxes,
        # Assign new face ID
```

```
else:
          box pt = left up pt, righ dow pt
          all_faces[curr_path][box_pt] = next_face_id
          next face id += 1
      # assign all the curr bounding boxes with distance
      # Larger than boundary
      unassigned indecies = curr bboxes indexes.difference(set(assigned indexies))
      for unassigned i in unassigned indecies:
        curr box = boxes[unassigned i]
        left up pt = curr box[0], curr box[1]
        righ dow pt = curr box[2], curr box[3]
        box_pt = left_up_pt, righ_dow_pt
        all_faces[curr_path][box_pt] = next_face_id
        next_face_id += 1
      bboxs prev = boxes
    else:
      all faces[curr path] = {}
      bboxs_prev = []
    prev path = curr path
    prev img = curr img
# Record time
end = time.time()
print("time takes: ", end - start)
```

#### time takes: 1.0520153045654297

```
In [18]:
          # Draw bounding boxes based on the tracks and save to some place
          save frames = {}
          for key in all_faces.keys():
            frame path = key
            curr img = cv2.imread(str(frame path))
            for box in all faces[key].keys():
              left_up_pt = box[0]
              righ dow pt = box[1]
              image drew box = cv2.rectangle(curr img,left up pt,
                        righ dow pt, (36,255,12), 1)
              cv2.putText(image_drew_box, str(all_faces[key][box]), (int(left_up_pt[0]),
                    int(left_up_pt[1])-10), cv2.FONT_HERSHEY_SIMPLEX, 0.8, (36,255,12), 2)
              save frames[frame path] = image drew box
          # Since nota all frames has a bounding box
          # here we save those frames without bounding box to save frames
          frames_with_bbox = set(save_frames.keys())
          for path in img paths:
            if str(path) not in frames with bbox:
              curr img = cv2.imread(str(path))
              save_frames[str(path)] = curr_img
```

```
In [19]: cv2_imshow(save_frames[list(save_frames.keys())[22]])
```



```
In [16]:
    outpath = './content/result_clip_3_method2'
    if not os.path.isdir(outpath):
        os.makedirs(outpath)
    for frame in save_frames:
            fname = os.path.split(frame)[-1]
            cv2.imwrite(os.path.join(outpath, fname),save_frames[frame])

In []:
    !ffmpeg -framerate 10 -pattern_type glob -i ./content/result_clip_3_method2/'*.jpg' \
            -vf "pad=ceil(iw/2)*2:ceil(ih/2)*2"\
            -c:v libx264 -pix_fmt yuv420p out3_method2.mp4
```