```
!pip3 install torch torchvision torchaudio pandas pyYAML tqdm seaborn opencv-pythor
Requirement already satisfied: torch in c:\users\jmess\appdata\local\programs\pyth
on\python39\lib\site-packages (1.12.0)
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Requirement already satisfied: kiwisolver>=1.0.1 in c:\users\jmess\appdata\local\p
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Requirement already satisfied: six>=1.5 in c:\users\jmess\appdata\local\programs\p
ython\python39\lib\site-packages (from python-dateutil>=2.8.1->pandas) (1.16.0)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\jmess\appdata\loc
al\programs\python\python39\lib\site-packages (from requests->torchvision) (1.26.1
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5)
Requirement already satisfied: idna<4,>=2.5 in c:\users\jmess\appdata\local\progra
ms\python\python39\lib\site-packages (from requests->torchvision) (3.3)
Requirement already satisfied: charset-normalizer<3,>=2 in c:\users\jmess\appdata
\local\programs\python\python39\lib\site-packages (from requests->torchvision) (2.
1.0)
```

In [ ]: %matplotlib inline

```
In [ ]: import torch
        import cv2
        from matplotlib import pyplot as plt
In [ ]: model = torch.hub.load('ultralytics/yolov5', 'yolov5s')
        Using cache found in C:\Users\jmess/.cache\torch\hub\ultralytics_yolov5_master
        YOLOv5 2022-7-18 Python-3.9.13 torch-1.12.0+cpu CPU
        Fusing layers...
        YOLOv5s summary: 213 layers, 7225885 parameters, 0 gradients
        Adding AutoShape...
In [ ]: def load_image(path):
             return cv2.imread(path)[:, :, ::-1]
In [ ]: def process_image(img):
            result = model([img], size=640)
             print(result.pandas().xyxy[0])
             return result
        IMGS_PATH = ['imgs/all.png', 'imgs/people_1.png', 'imgs/people_2.png']
In [ ]: |
In [ ]: IMGS = []
        for i in IMGS_PATH:
            IMGS.append(load_image(i))
        a)
In [ ]: def draw_result(frame, results):
             frame = frame.copy()
             for box in results.xyxy[0]:
                 if int(box[5]) == 0:
                     xB = int(box[2])
                     xA = int(box[0])
                     yB = int(box[3])
                    yA = int(box[1])
                     cv2.rectangle(frame, (xA, yA), (xB, yB), (255, 0, 0), 2)
             return frame
In [ ]: def process(img):
             plt.imshow(img)
             plt.show()
             result = process_image(img)
             print(result)
             plt.imshow(draw_result(img, result))
             plt.show()
```

In [ ]: process(IMGS[0])

```
200 - 400 - 600 - 1200 - 1250 1500 1750
```

	xmin	ymin	xmax	ymax	confidence	class	\
0	726.216492	870.891785	967.580078	1088.871460	0.846512	2	
1	1104.301514	841.032959	1196.471558	923.366333	0.829974	2	
2	393.870117	922.003052	694.894958	1192.261230	0.800172	2	
3	1.707250	910.309021	84.191330	1184.690308	0.776526	0	
4	636.185120	826.378418	760.100464	929.397766	0.770256	2	
5	1045.868408	804.348999	1140.609253	883.489014	0.762862	2	
6	1600.351807	493.414764	1672.191406	587.836914	0.680482	9	
7	256.838928	852.810669	328.951660	1052.742432	0.631074	0	
8	447.830383	725.431458	625.831116	910.149231	0.628749	5	
9	1189.572998	697.095337	1907.240234	1186.575684	0.597191	7	
10	1206.703735	593.303833	1252.806030	651.825256	0.585567	9	
11	878.548706	793.447327	911.642090	870.598999	0.522277	0	
12	971.822571	797.682190	1003.063477	869.149353	0.472607	0	
13	718.716797	753.658752	812.098267	850.487915	0.451698	2	
14	396.205566	838.769409	441.134613	986.398010	0.450832	0	
15	507.221191	523.475037	573.016846	602.472778	0.437403	9	
16	1020.958191	670.210266	1041.742432	708.367859	0.422663	9	
17	392.386963	924.273560	691.774841	1186.112305	0.386347	7	
18	1196.590454	700.699097	1843.417358	1188.745605	0.296920	5	
19	964.837463	767.659180	1010.002258	861.787720	0.285839	0	

```
name
0
               car
1
               car
2
               car
3
            person
4
5
               car
6
    traffic light
7
            person
8
               bus
9
             truck
10
    traffic light
11
            person
12
            person
13
               car
14
            person
15
    traffic light
16
    traffic light
17
             truck
18
               bus
            person
```

image 1/1: 1235x1920 6 persons, 6 cars, 2 buss, 2 trucks, 4 traffic lights
Speed: 14.9ms pre-process, 116.6ms inference, 16.9ms NMS per image at shape (1, 3, 416, 640)



Como podemos ver, já que existem varios elementos na foto boa parte das pessoas foi encontrada pelo modelo. Mas dá para ver que quanto mais profundo na foto o elemento está menor a confiança na classificação temos. Provavelmente devido menor informação de pixels para representá-lo dentro da foto. A rede identificou todas as pessoas na foto sem nenhuma falha.

## In [ ]: process(IMGS[1])

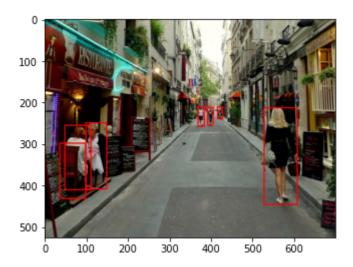


20/07/2022 12:57 segunda questao

```
xmin
                                                   confidence class
                     ymin
                                 xmax
                                             ymax
   528.242798 213.827530
                           608.271057
                                       445.049377
                                                     0.859260
                                                                   0
1
   369.474243
               211.207169
                           384.192078
                                       257.956268
                                                     0.816155
                                                                   0
2
    392.125702
               212.672897
                           405.924194
                                       254.231247
                                                     0.733099
                                                                   0
3
    99.137474
               249.691849
                           149.096802
                                       408.160004
                                                     0.711861
                                                                   0
4
   519.139893 276.049927
                           558.115784
                                       353.744110
                                                     0.657766
                                                                  26
5
    48.602715 255.429077
                                                                   0
                           111.330231 410.260864
                                                     0.655491
6
   445.193054 199.727753 473.608276 257.162567
                                                     0.448671
                                                                  58
7
   582.135010
               246.941681 621.685974
                                       377.284088
                                                     0.395870
                                                                  58
8
   416.761749
               210.722229
                          428.724884
                                       239.429962
                                                     0.371667
                                                                   0
9
    34.038357
               305.115082
                            92.542786
                                       430.863617
                                                     0.362988
                                                                  56
10
                           617.694824
                                                                  58
   582.520630
               325.183167
                                       377.309143
                                                     0.335677
   659.666870 109.544678 700.000000 153.703491
                                                                  58
                                                     0.307373
   521.921143 209.655411 605.688599 448.508728
                                                     0.261612
                                                                  26
13
    36.373814 298.593567
                            98.360062 431.164307
                                                     0.258987
                                                                   0
```

```
name
0
           person
1
           person
2
           person
3
           person
4
         handbag
5
           person
6
    potted plant
7
    potted plant
8
           person
9
            chair
10
    potted plant
    potted plant
11
12
         handbag
           person
```

image 1/1: 525x700 7 persons, 2 handbags, 1 chair, 4 potted plants
Speed: 4.0ms pre-process, 121.6ms inference, 1.0ms NMS per image at shape (1, 3, 4
80, 640)



Uma rua com uma quantidade relativa de objetos podemos ver que todas as pessoas foram identificadas, contudo o modelo acabou errado e identificado uma pessoa duas vezes.

```
In [ ]: process(IMGS[2])
```

```
0
 250
 500
 750
1000
1250
1500
1750
              500
                       1000
                                 1500
                                           2000
                                                     2500
     0
            xmin
                            ymin
                                             xmax
```

```
    xmin
    ymin
    xmax
    ymax
    confidence
    class

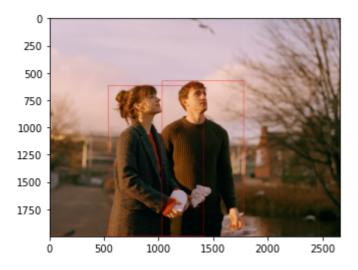
    0
    1028.121338
    574.823120
    1774.388672
    1998.000000
    0.915181
    0

    1
    537.640442
    618.084961
    1406.829224
    1992.961426
    0.865246
    0
```

name 0 person 1 person

image 1/1: 1998x2664 2 persons

Speed: 24.9ms pre-process, 111.6ms inference, 1.0ms NMS per image at shape (1, 3, 480, 640)



Os dois unico elementos da foto propositalmente são duas pessoas e podemos ver que o modelo o identifica, com um nivel consideravel de confiança as duas pessoas.

## b)

```
In []: # Transforma o video em imagens para ser usado no modelo

def get_frames(path):
    vidcap = cv2.VideoCapture(path)
    vidcap.set(cv2.CAP_PROP_FRAME_WIDTH, 1280)
    vidcap.set(cv2.CAP_PROP_FRAME_HEIGHT, 720)
    lista_results = []

    success,image = vidcap.read()
    while success:
        image = cv2.cvtColor(image, cv2.COLOR_BGR2RGB)
        lista_results.append(image)
```

```
success,image = vidcap.read()
            return lista_results
In [ ]: # Pega todos os frames
        results = get_frames('videos/cross.mp4')
In [ ]: # Executa os frames no modelo
        t = model(results)
        pd = t.pandas().xyxy
In [ ]: # Verifica e calcula os centros dos quadrados demarcadores
        def get_path(pd):
            path = []
            for j in pd:
                i = j[j['class'] == 0]
                if len(i) > 0:
                     i = i.iloc[0]
                     x = (i['xmax'] + i['xmin']) // 2
                    y = (i['ymax'] + i['ymin']) // 2
                    path.append((int(x), int(y)))
            return path
In [ ]: # Pinta aréa a partir dos centros dos quadrados demarcadores calculado
        def print_area(img, point, w):
            for i in range(point[0] - w, point[0] + w):
                for j in range(point[1] - w, point[1] + w):
                     try:
                         img[j][i] = (255, 0, 0)
                     except IndexError:
                        pass
In [ ]: # Angraria os centros
        a = get_path(pd)
In [ ]: # Pinta o caminho na imagem
        img = results[0].copy()
        for i in a:
            print_area(img, i, 30)
In [ ]: # Exibe a imagem
        plt.imshow(img)
        plt.show()
```

