```
!pip3 install torch torchvision torchaudio pandas pyYAML tqdm seaborn opencv-pythor
Requirement already satisfied: torch in c:\users\jmess\appdata\local\programs\pyth
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Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\jmess\appdata\loc
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1.0)
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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)
```

In []: %matplotlib inline

```
import torch
        import cv2
        from matplotlib import pyplot as plt
        import numpy as np
        from IPython.display import Image, display
        from io import StringIO
        import PIL.Image
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/bus.png', 'imgs/people.png']
In [ ]:
In [ ]: IMGS = []
```

a)

```
In [ ]: plt.imshow(IMGS[0])
    plt.show()
```

for i in IMGS_PATH:

IMGS.append(load_image(i))



```
In [ ]: process_image(IMGS[0])
```

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

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

Out[]: <models.common.Detections at 0x2cf7ff73bb0>

Como podemos ver, já que existem varios elementos na foto boa parte deles foi encontrado 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.

```
In [ ]: plt.imshow(IMGS[1])
   plt.show()
```



```
In []: process_image(IMGS[1])

xmin ymin xmax ymax confidence class name
0 35.439453 164.285309 1411.653564 1055.357544 0.862536 5 bus

Out[]: cmodels.common.Detections at 0x2cf7ff73730>
```

O unico elementos da foto propositalmente é um onibus e podemos ver que o modelo o identifica, com um nivel consideravel de confiança.

```
In [ ]: plt.imshow(IMGS[2])
   plt.show()
```



```
process_image(IMGS[2])
In [ ]:
                                                          ymax
                                                                confidence
                                                                             class
                   xmin
                               ymin
                                             xmax
            1028.121338
                                     1774.388672
                        574.823120
                                                   1998.000000
                                                                  0.915181
                                                                                 0
             537.640442
                        618.084961
                                     1406.829224
                                                   1992.961426
                                                                  0.865246
                                                                                 0
              name
           person
            person
         <models.common.Detections at 0x2cf07326bb0>
Out[]:
```

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.

```
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
        # 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):
                         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()

