Project 2 - PII Detection Using YOLO

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1. Installed darknet, downloaded the pre-trained weight file and finished setting up yolo.

2. Testing Yolo

```
512 1 x 1 / 1
1024 3 x 3 / 1
                                                                                               0.379 BFLOPs
                                           19 x
                                                  19 x1024
                                          19 x 19 x 512
                                                                          19 x 19 x1024
                                                                                               3.407 BELOPS
               71
512
                                          19 x
                                                  19 x1024
                                                                          19 x 19 x1024
                 1

512 1 x 1 / 1

1024 3 x 3 / 1

512 1 x 1 / 1

1024 3 x 3 / 1

512 1 x 1 / 1

1024 3 x 3 / 1

255 1 x 1 / 1
                                          19 x 19 x1024
   75 conv
                                                                                               0.379 BFLOPs
                                                                          19 x 19 x1024
                                                                                               3.407 BFLOPs
                 1024
                                           19 x
                                                                                               0.379 BFLOPs
                                           19 x 19 x1024
                                                                          19 x 19 x 512
   77 conv
                                                                          19 x 19 x1024
19 x 19 x 512
   78 conv
                 1024
                                           19 x
                                                                                               3.407 BFLOPs
                                                  19 x1024
                                                                                               0.379 BFLOPs
   79 conv
                                          19 x
                                                                         19 x 19 x1024
19 x 19 x 255
                                                  19 x 512
19 x1024
   80 conv
                 1024
                                                                                               3.407 BFLOPs
                                                                                               0.189 BFLOPs
   81 conv
   82 yolo
   83 route
                                                                         19 x 19 x 256
38 x 38 x 256
                                                                                               0.095 BFLOPs
   84 conv
                                          19 x 19 x 256
   85 upsample
                85 61
   86 route
                 35 61

256 1 x 1 / 1

512 3 x 3 / 1

256 1 x 1 / 1

512 3 x 3 / 1

256 1 x 1 / 1

512 3 x 3 / 1

255 1 x 1 / 1
                                                                         38 x 38 x 256
38 x 38 x 512
   87 conv
                                          38 x
                                                                                               0.568 BFLOPs
                                          38 x 38 x 256
                                                                                               3.407 BFLOPs
   88 conv
                                          38 x 38 x 512
                                                                         38 x 38 x 256
38 x 38 x 512
   89 conv
                                                                                               0.379 BFLOPs
   90 conv
                                          38 x
                                                  38 x 256
                                                                                               3.407 BFLOPs
   91 conv
                                                  38 x 512
                                                                                 38 x 256
38 x 512
                                          38 x
                                                                          38 x
                                                                                               0.379 BFLOPs
                                                                                               3.407 BFLOPs
   92 conv
                                          38 x
                                                  38 x 256
                                                                          38 x
                                                                          38 x 38 x 255
                                                                                               0.377 BFLOPs
                                          38 x 38 x 512
   93 conv
   94 yolo
   95 route
                91
                                                                         38 x 38 x 128 0.095 BFLOPs
76 x 76 x 128
                 128 1 x 1 / 1
e 2x
                                          38 x 38 x 256
38 x 38 x 128
   96 conv
   97 upsample
                97 36
   98 route
   99 conv
                                          76 x 76 x 384
                                                                         76 x 76 x 128
                                                                                               0.568 BFLOPs
  100 conv
                                          76 x
                                                                                               3.407 BFLOPs
                  256
                                                                         76 x
                  128 1 x 1 / 1
256 3 x 3 / 1
128 1 x 1 / 1
256 3 x 3 / 1
255 1 x 1 / 1
                                          76 x
                                                  76 x 256
                                                                         76 x 76 x 128
                                                                                               0.379 BFLOPs
  101 conv
                                                                         76 x 76 x 256
76 x 76 x 128
                                                  76 x 128
                                                                                               3.407 BFLOPs
  102 conv
                                          76 x
                                                                                              0.379 BFLOPs
                                          76 x 76 x 256
  103 conv
  104 conv
                                          76 x 76 x 128
76 x 76 x 256
                                                                         76 x 76 x 256 3.407 BFLOPs
76 x 76 x 255 0.754 BFLOPs
  105 conv
  106 yolo
Loading weights from yolov3.weights...Done!
data/dog.jpg: Predicted in 46.715687 seconds.
dog: 100%
truck: 92%
bicycle: 99%
```

3. Downloaded and extracted dataset. The dataset has 109 images in "private" folder and 100 images in "public" folder.

```
@LAPTOP-7EFF46BN:~/photoprivacy/darknet$ unzip dataset.zip
Archive: dataset.zip
  creating: dataset/private/
 inflating: dataset/private/4330400088.jpg
 inflating: dataset/private/4331808437.jpg
 inflating: dataset/private/4333240960.jpg
 inflating: dataset/private/4333681728.jpg
 inflating: dataset/private/4333696292.jpg
 inflating: dataset/private/4336675220.jpg
 extracting: dataset/private/4337733437.jpg
 inflating: dataset/private/4338158943.jpg
 inflating: dataset/private/4338398780.jpg
 extracting: dataset/private/4338422575.jpg
 inflating: dataset/private/4338457809.jpg
 inflating: dataset/private/4338666198.jpg
 extracting: dataset/private/4339315528.jpg
 extracting: dataset/private/4339819203.jpg
 extracting: dataset/private/4340780316.jpg
 inflating: dataset/private/4342990075.jpg
 inflating: dataset/private/4343147011.jpg
 inflating: dataset/private/4343170577.jpg
 extracting: dataset/private/4344397460.jpg
 inflating: dataset/private/4344531492.jpg
 inflating: dataset/private/4344715961.jpg
 inflating: dataset/private/4345628858.jpg
 inflating: dataset/private/4347435013.jpg
 inflating: dataset/private/4347738656.jpg
 inflating: dataset/private/4348957203.jpg
 inflating: dataset/private/4349297203.jpg
 inflating: dataset/private/4350442802.jpg
 extracting: dataset/private/4350870822.jpg
 inflating: dataset/private/4351924993.jpg
 inflating: dataset/private/4352674310.jpg
 inflating: dataset/private/4353117303.jpg
 inflating: dataset/private/4353135983.jpg
 inflating: dataset/private/4353651367.jpg
 inflating: dataset/private/4353808437.jpg
 extracting: dataset/private/4353816194.jpg
 inflating: dataset/private/4353934580.jpg
 inflating: dataset/private/4354000011.jpg
 extracting: dataset/private/4354219152.jpg
 inflating: dataset/private/4355569364.jpg
```

4. Results of running detectors – "public.py" and "private.py" on public and private datasets are stored in "public.txt" and "private.txt" respectively.

```
krupa@LAPTOP-7EFF46BN: ~/photoprivacy/darknet
                                            krupa@LAPTOP-7EFF46BN: ~/photoprivacy/darknet
diningtable: 60%
hot dog: 57%
                                            person: 100%
                                            person: 100%
chair: 83%
                                            person: 100%
person: 99%
                                           person: 98%
person: 91%
person: 63%
                                           person: 100%
                                           person: 96%
person: 100%
person: 55%
                                            person: 73%
                                            person: 100%
person: 99%
person: 99%
                                           dog: 75%
person: 97%
                                            person: 92%
person: 95%
person: 86%
                                            person: 100%
                                            person: 99%
chair: 72%
chair: 68%
                                            cup: 86%
chair: 62%
                                            cup: 53%
cup: 92%
                                            person: 100%
bottle: 70%
                                            person: 55%
person: 99%
                                            person: 55%
person: 97%
                                            person: 53%
person: 95%
person: 84%
                                            person: 100%
person: 81%
person: 79%
                                            person: 100%
person: 74%
person: 70%
                                           cup: 89%
person: 59%
                                            cup: 62%
                                            person: 100%
person: 59%
person: 100%
person: 99%
                                           book: 65%
person: 99%
                                           microwave: 52%
person: 99%
                                            cell phone: 65%
person: 98%
person: 96%
                                           person: 100%
person: 92%
"public.txt" 293L, 2401C
                                            "private.txt" 381L, 3411C
```

5. Results of running "results.py" to obtain top 10 images in both public and private category

```
upa@LAPTOP-7EFF46BN:<mark>~/photoprivacy/darknet$ python3 results.p</mark>y
private.txt
Num.
        Occurances
                         Object
        156
                         person
                         cup
        8
                         dog
        8
                         wine glass
                         chair
                         cell phone
                         tymonitor
                         diningtable
                         bottle
10.
                         book
public.txt
        Occurances
Num.
                         Object
        124
                         person
                         bottle
                         chair
                         knife
                         giraffe
                         book
                         train
                         diningtable
10.
                         cup
crupa@LAPTOP-7EFF46BN:~/photoprivacy/darknet$
```

Observations

- 1. 'Person' is the top result for both private and public images dataset. This means, people tend to keep their photos as both public and private.
- 2. Some objects such as "book", "cup" and "chair" fall under both the categories indicating that the environment in which these object are placed decides the category they have to fall under.
- 3. Usually those objects which are inaccessible by all and are confined to one place fall under private category and those which can be accessed by general public fall under public category.
- 4. Personal belongings example, "wine glass", "cell phone" and "tymonitor" fall under private.

Appendix

CODES:

private.py

```
import os
from subprocess import Popen, PIPE
file1 = open("private.txt", "a+", encoding = "utf-8")
images_dir = '/home/krupa/photoprivacy/darknet/dataset'
setting = 'private'
for img in os.listdir(os.getcwd()+"/dataset/private"):
  count =0
  p = Popen(['./darknet', 'detect', 'cfg/yolov3.cfg', 'yolov3.weights', images_dir + '/' + setting + '/' +
  str(img)],cwd = '/home/krupa/photoprivacy/darknet', stdout = PIPE, stderr = PIPE)
  stdout, stderr = p.communicate()
  output = stdout.decode("utf-8")
  output_split = output.split("\n")
  for i in output_split:
    if "Predicted in" not in i:
       print(i)
      file1.write(i)
      file1.write("\n")
file1.close()
```

public.py

```
import os
from subprocess import Popen, PIPE
file1 = open("public.txt", "a+", encoding = "utf-8")
images_dir = '/home/krupa/photoprivacy/darknet/dataset'
setting = 'public'
for img in os.listdir(os.getcwd()+"/dataset/public"):
  count =0
  p = Popen(['./darknet', 'detect', 'cfg/yolov3.cfg', 'yolov3.weights', images_dir + '/' + setting + '/' +
  str(img)], cwd = '/home/krupa/photoprivacy/darknet', stdout = PIPE, stderr = PIPE)
  stdout, stderr = p.communicate()
  output = stdout.decode("utf-8")
  output_split = output.split("\n")
  for i in output_split:
    if "Predicted in" not in i:
       print(i)
      file1.write(i)
      file1.write("\n")
file1.close()
```

results.py

```
import string
import os
file_list = ['private.txt','public.txt']
for file_name in file_list:
  d = dict()
  contents = []
  print("\n")
  print(file_name)
  print("*********")
  file_name = open (file_name, "r", encoding = 'utf-8')
  lines_private=[line.rstrip('\n') for line in (file_name)]
  lines_private = list(filter(None, lines_private))
  for line in lines_private:
    index = line.find(":")
    value = line[0:index]
    contents.append(value)
  for line in contents:
    if line in d:
       d[line] = d[line] + 1
    else:
       d[line] = 1
  sorted_d = sorted(d, key=d.get, reverse=True)
  #print(sorted_d)
  print("{}.\t{}\t{}\".format("Num", "Occurances", "Object"))
  for i in range(10):
    #print(i+1,sorted_d[i], d[sorted_d[i]])
    print("{}.\t{}\t\t{}".format(i+1,d[sorted_d[i]], sorted_d[i]))
  print("\n")
  file_name.close()
```