Hardware II Seminar

Sensors and Data Analysis





Quick recap



Image and Video processing

Interfacing with cameras



Basics of image processing



Image and Video processing - Recap





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Video capture and storage of RGB camera example:

```
import cv2
                                            python
cap = cv2.VideoCapture(0)
if (cap.isOpened() == False):
     print("Unable to read camera feed")
frame width = int(cap.get(3))
frame height = int(cap.get(4))
cv2.VideoWriter('out.avi',cv2.VideoWriter fourcc('M'
','J','P','G'), 10, (frame width, frame height))
while(cap.isOpened()):
    ret, frame = cap.read()
    if ret == True:
        frame = cv2.flip(frame, 0)
        out.write(frame)
        cv2.imshow('frame',frame)
        if cv2.waitKey(1) & 0xFF == ord('q'):
            break
    else:
        break
cap.release()
out.release()
cv2.destroyAllWindows()
```



Open video capture on device 0 (webcam in my case)

VideoWriter (see https://www.fourcc.org/fourcc.php)

Capture, store and show frame

Close everything if q is pressed





Image and Video processing - Some use cases







Here are some of the possible use cases:





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- <u>Feature detection</u>: descriptors techniques
- <u>Image alignment</u>: align images in pictures, homography
- <u>Object detection</u>: detect object in images (where something is, knowing what we are looking for) solvable with DL, but ML works too
- <u>Identification</u>: identify if one object matches another





Understanding the basics

Descriptors



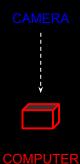












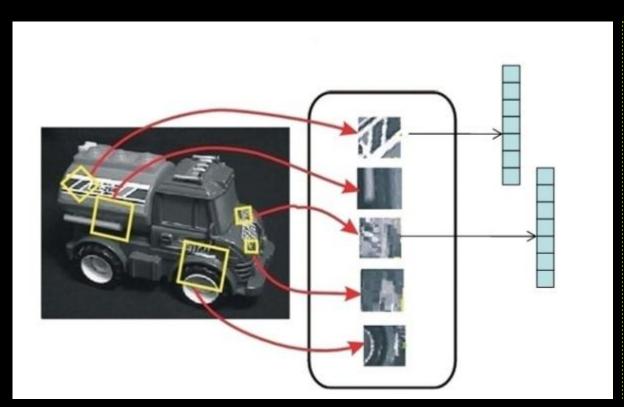
When dealing with image processing problems, a common approach is to divide the image in **subimages or patches**. Each of the patches is a small portion of the original image which can then be used and compared to other patches of other images and extract conclusions. Are the objects the same? Are the objects of the same type? Ideally, we would need of a method to determine if a patch is the same to another one, and that this method is independent of image changes: rotation, lighting, noise... Welcome **descriptors**.











> A descriptor is some function that is applied on the patch to describe it in a way that is invariant to all the image changes that are suitable to our application (e.g. rotation, illumination, noise etc.). A descriptor is "built-in" with a distance function to determine the similarity, or distance, of two computed descriptors. So to compare two image patches, we'll compute their descriptors and measure their similarity by measuring the descriptor similarity, which in turn is done by computing their descriptor distance.









Some descriptor types

- HOG: Histogram of Oriented Gradients. Nice wrap-up <u>here</u>
 - SIFT (non free)
 - SURF (non free)
- <u>Binary descriptors</u>: Are a way of representing a patch as a binary string, using only comparison of intensity (in separated channels if necessary)
 - BRIEF
 - ORB
 - BRISK
 - FREAK
 - 0 ...

They are based on the following workflow. 1. Sampling pattern definition: where to sample points in the region around the descriptor. 2. Orientation compensation: some mechanism to measure the orientation of the keypoint and rotate it to compensate for rotation changes. 3. Sampling pairs: the pairs to compare when building the final descriptor.







Basic feature finding example



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```
#!/usr/bin/python
                                  python
import numpy as np
import cv2 as cv
from matplotlib import pyplot as plt
img = cv.imread('images/eagle.jpg',0)
img = cv.GaussianBlur(img, (3, 3), 0)
# Initiate ORB detector
orb = cv.ORB create()
# find the keypoints with ORB
kp = orb.detect(img, None)
# compute the descriptors with ORB
kp, des = orb.compute(img, kp)
# draw only keypoints location, not size
and orientation
img2 = cv.drawKeypoints(img, kp, None,
color=(0,255,0), flags=0)
cv.imshow('Features', img2)
cv.imwrite("images/eaglefeatures.jpg",
```



Open image and filter it

Initiate ORB and detect keypoint

Show result keypoings, and save







Matching features in two images

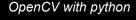




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```
#!/usr/bin/python
                                    python
import numpy as np
import cv2 as cv
import matplotlib.pyplot as plt
imq1 =
cv.imread('images/eaglerot.jpg',cv.IMREAD
GRAYSCALE)
img1 = cv.GaussianBlur(img1, (3,3), 0)
orb = cv.ORB create()
kp1 = orb.detect(img1, None)
kp1, des1 = orb.compute(imq1, kp1)
bf = cv.BFMatcher(cv.NORM HAMMING,
crossCheck=True)
matches = bf.match(des1, des2)
matches = sorted(matches, key = lambda
x:x.distance)
imq3 =
cv.drawMatches(img1, kp1, img2, kp2, matches[:
[10], None, flags=cv.DrawMatchesFlags NOT DRA
W SINGLE POINTS)
plt.imshow(imq3),plt.show()
```







Open image 1 (and also image 2)

Initiate ORB and detect keypoints in both images

Get matches. Sort them in the order of their distance and draw them









Recommended content

- OpenCV Documentation: <u>https://docs.opencv.org/3.4/db/d27/tutorial_py_table_of_contents_feature2d.html</u>
- Learn OpenCV: https://learnopencv.com
- GilsCV Blog: https://gilscvblog.com/

> Try to understand well these basics to know what you are doing!





Some techniques

Homography





Image and Video processing - Some techniques





Homography is a transformation (a matrix) that maps the points in one image to the corresponding points in the other image.

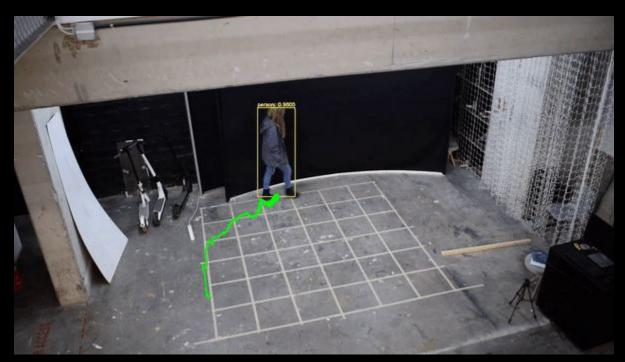






Image and Video processing - Some techniques





Homography is a transformation (a matrix) that maps the points in one image to the corresponding points in the other image

Some examples:

https://www.learnopencv.com/homo graphy-examples-using-opencv-pyt hon-c/

https://www.learnopencv.com/imag e-alignment-feature-based-using-op encv-c-python/







Some techniques

Tracking things



Image and Video processing - Tracking things

Ultimately, we can try to detect things in a image, or video frame (the same thing). We can <u>track</u> objects in an image by <u>tracking</u> color, pixel changes, or trying to build more complex trackers, all the way up to deep learning algorithms.

Remember, tracking, in general, doesn't imply identifying what the tracked object is, so normally it's a more light weight process.







Source:

https://docs.opencv.org/trunk/da/d97/tutorial_threshold_inRange.html

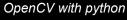
Image and Video processing - Color tracking



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```
#!/usr/bin/python
                                            python
import cv2 as cv
cap = cv.VideoCapture(args.camera)
while True:
   ret, frame = cap.read()
   if frame is None:
break
   frame HSV = cv.cvtColor(frame, cv.COLOR BGR2HSV)
    frame threshold = cv.inRange(frame HSV, (low H,
low S, low V), (high H, high S, high V))
   bn img = cv.erode
(frame threshold, cv.getStructuringElement(cv.MORPH RE
CT, (3,3)), iterations = 1)
   bn img = cv.dilate
(bn img, cv.getStructuringElement(cv.MORPH RECT, (5,5))
,iterations = 1)
   M = cv.moments(bn img)
   if M['m00']>50000:
       cx = int(M['m10']/M['m00'])
       cy = int(M['m01']/M['m00'])
       cv.circle (frame, (cx, cy), 20, (255, 0, 0), 2)
    cv.imshow(window capture name, frame)
   cv.imshow(window detection name, frame threshold)
    kev = cv.waitKev(30)
   if key == ord('q') or key == 27:
        break
```







Open video and start loop

Convert to HSV, get pixels within Range. Clean Image

Get moments, centroid and draw a circle. Then show

Full example

https://github.com/oscgonfer/sensors_dsp_lectures/blob/cu rrent/03 computer vision/examples/06 color tracking.py





Image and Video processing - Color tracking / use cases

For an image, we get all the pixels that fall between a certain range of HSV or RGB values.

Then, we calculate the centroid of those coordinates.

This technique is only useful if there are very distinct colors, but is very fast if images are not <u>super-mega high-resolution</u>.







Source:

https://docs.opencv.org/4.2.0/d4/dee/tutorial_optical_flo

Image and Video processing - Optical Flow

Optical flow or optic flow is the pattern of apparent motion of objects, surfaces, and edges in a visual scene caused by the relative motion between an observer and a scene.





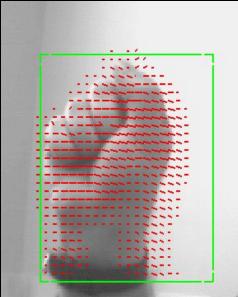






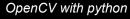


Image and Video processing - Optical Flow



```
#!/usr/bin/python
                                               python
import numpy as np
import cv2 as cv
cap = cv.VideoCapture(0)
ret, old frame = cap.read()
old gray = cv.cvtColor(old frame, cv.COLOR BGR2GRAY)
p0 = cv.goodFeaturesToTrack(old gray, mask = None,
**feature params)
while (1):
    ret, frame = cap.read()
    frame gray = cv.cvtColor(frame,
cv.COLOR BGR2GRAY)
    p1, st, err = cv.calcOpticalFlowPyrLK(old gray,
frame gray, p0, None, **lk params)
    if p1 is not None:
      good new = p1[st==1]
      good old = p0[st==1]
      for i, (new, old) in enumerate (zip (good new,
good old)):
          a,b = new.ravel()
          c,d = old.ravel()
          mask = cv.line(mask, (a,b), (c,d),
color[i].tolist(), 2)
            frame =
cv.circle(frame, (a,b),5,color[i].tolist(),-1)
        img = cv.add(frame, mask)
        cv.imshow('frame',img)
        old gray = frame gray.copy()
        p0 = good new.reshape(-1,1,2)
```







Open video and get features to track

Get optical flow and get good points.

Draw lines

Replace new frame and start over

Full example:

https://github.com/oscgonfer/sensors_dsp_lectures/blob/current/03 _computer_vision/examples/08_optical_flow_lk.py





Some techniques

Identifying things





(a) Edge Features

(b) Line Features

(c) Four-rectangle features

https://docs.opencv.org/3.4/db/d28/tutorial_cascade_classifier

Image and Video processing - Identifying things





When identifying things, it's better to try to find classifiers that try to find something very specific.

The more specific, the faster it normally will be, for instance, using ML technique of Cascade Classifiers.

> Object Detection using Haar feature-based cascade classifiers is an effective object detection method proposed by <u>Paul Viola and Michael Jones</u> in their paper, "Rapid Object Detection using a Boosted Cascade of Simple Features" in 2001. It is a machine learning based approach where a cascade function is trained from a lot of positive and negative images. It is then used to detect objects in other images.

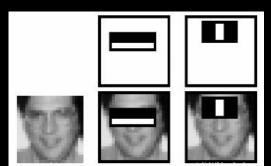
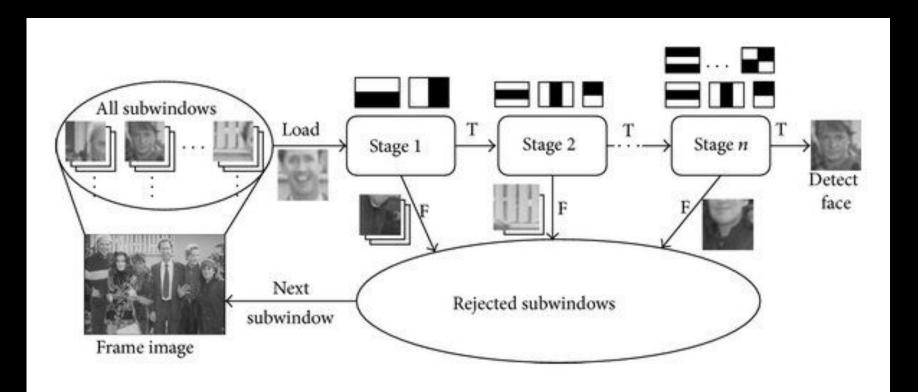






Image and Video processing - Haar Cascade Classifier







Taac

Image and Video processing - Identifying things









Image and Video processing - Identifying things









Source:

https://docs.opencv.org/3.4/db/d28/tutorial_cascade_classifier

Image and Video processing - Cascade classifier

python



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Full example:

https://github.com/oscgonfer/sensors dsp_lectures/blob/current/03_comput er_vision/examples/07_classifier.py











Some techniques

Identifying things - the DL way









> YOLO performs object detection by creating a rectangular grid throughout the entire image. Then creates bounding boxes based on (x,y) coordinates. Class probability gets mapped by using random color assignment. To filter out weak detections, a 50% confidence is being used (this can change) which helps eliminate unnecessary boxes.

The paper: https://arxiv.org/abs/1804.02767
The source: https://pireddie.com/darknet/yolo/









Installation:

- Install Darknet: https://pjreddie.com/darknet/install/
 - a. For weird image formats, use it with OPENCV (optional). For this, modify Makefile.
- 2. If using opency and opency-python, it comes already with darknet

Weights and cfg (remember, this is DL)

Get the **weights** from <u>here</u>:

```
wget https://pjreddie.com/media/files/yolov3.weights
```

Get the yolov3.cfg from here:

wget https://github.com/pjreddie/darknet/blob/master/cfg/yolov3.cfg

Check in the **Darknet** github for more cfgs.

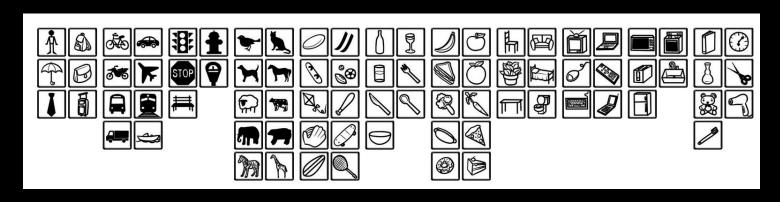






COCO dataset (COmmon objects in COntext) - https://cocodataset.org/

COCO dataset is trained on 200,000 images of 80 different categories (person, suitcase, umbrella, horse etc...). It contains images, bounding boxes and 80 labels. COCO can be used for both object detection, and segmentation, we are using it for detecting people.



We will be using the COCO dataset. You will find coco.names in the extras folder of the repository





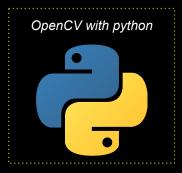


Full image example:

https://github.com/oscgonfer/s
ensors_dsp_lectures/blob/curr
ent/03_computer_vision/exam
ples/09_yolo.py

Full video example:

https://github.com/oscgonfer/s
ensors_dsp_lectures/blob/curr
ent/03_computer_vision/exam
ples/10_volo_video.py







References

- COCO Common Objects in Context
- <u>Learning Computer Vision Basics in Excel</u>
- Corner Detector
- Real time person removal
- <u>Introduction to descriptors</u> and <u>Tutorial on descriptors</u>
- GilCV Code Examples
- Viola-Jones face detection and tracking explained
- <u>Using non-free detectors</u>





References - Working on a pi

- Rpi camera do's and dont's
- Optimizing opency for the pi
- Face Recognition
- Object Detection with DL
- Facial landmarks







Thanks!



