

Computer Vision Assignment Two @ ETH Zurich

Feature Extraction

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1 Introduction

Feature extraction is used in many application in Computer Vision such as motion detection, visual object tracking, image reconstruction..etc.

In this assignment we implement the harris corner detector and then we perform feature matching. The results obtained are then compared to the SIFT detector.

2 Feature Extraction

First we compute the gradients in x and y direction using the built-in function in matlab that takes care implicitly for the boundary cases in the image. Afterwards, the image is smoothed using a gaussian filter and the components needed for the harris corner detector are calculated which are then used to calculate the response measure. Through plotting the histogram of the response measure of all the image pixels; an initial estimate of the threshold is observed.

$$H = \begin{bmatrix} I_x^2 & I_x * I_y \\ I_x * I_y & I_y^2 \end{bmatrix}$$

$$K = \det(H)/\text{trace}(H) = (I_x^2 * I_y^2 - (I_x * I_y)^2)/(I_x^2 + I_y^2 + \text{eps})$$

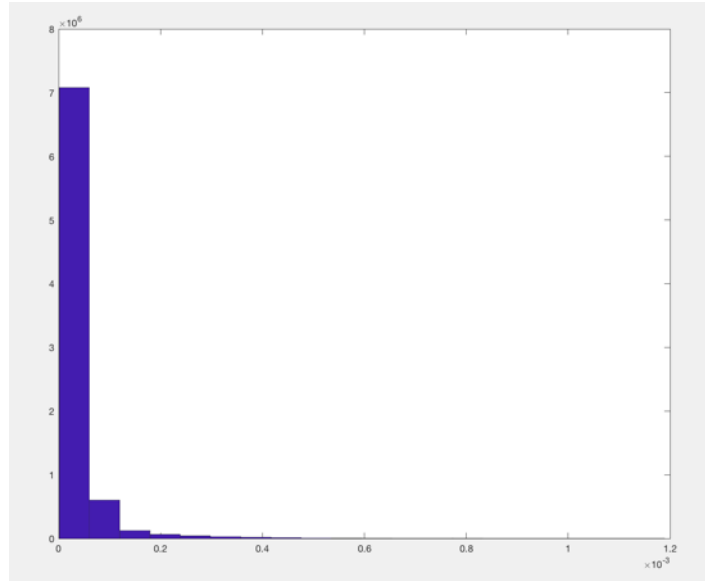


Figure 1: Harris Corner Response

Afterwards, non max suppression is applied to the K global matrix without removing the response measures that are below the threshold (for faster performance). To apply the non max suppression on the whole matrix, `ordfilt2` was used and a 3×3 window.

To determine the final keypoint features. A boolean expression is used.
 $KeyPoints = (nonmax == K) \& K > threshold$

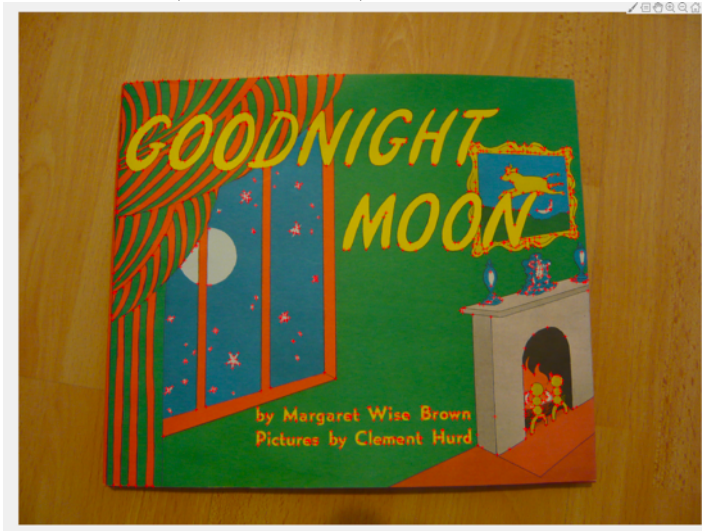


Figure 2: Corners using Harris Detector in Image 1

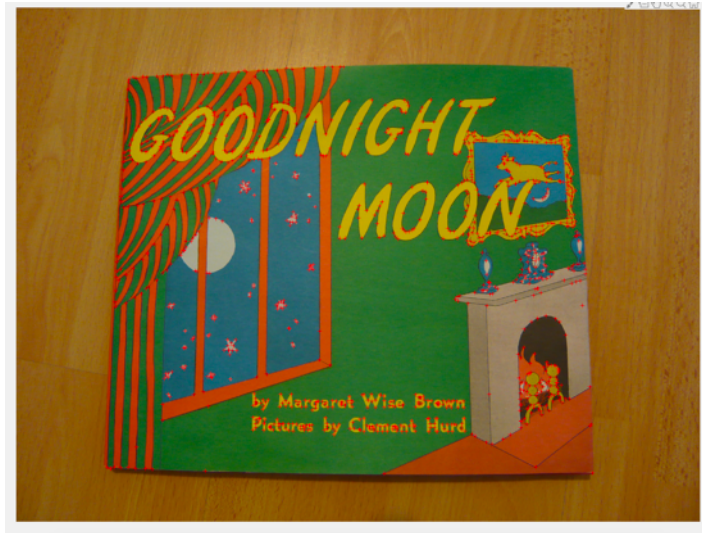


Figure 3: Corners using Harris Detector in Image 2

Number of corners in Image 1 = 886

Number of corners in Image 2 = 1122

3 Feature Descriptor

For each key point, we save the patch around it as a descriptor in a 3x3 window, so we can compare the two descriptors available from the two images to deduce the matches.

4 SSD Feature Matching

For each key point descriptor in one image, we detect if there is a corresponding key point in the other image where the squared difference is really small that we can be confident that it's an actual match of the point. The threshold used to define how small the square difference should be so that it can be accepted as a match, was calculated using trial and error. Small differences denote high similarity.

Total Number of Matches = 224



Figure 4: Feature Matching using the Harris Corner Detection Results.

5 SIFT Features using VLfeat

The tutorial to use VLFeat (SIFT feature extractor from Andrea Vedaldi) was followed to apply SIFT on the two images provided. The features were extracted using SIFT detector and then matched. To decrease the number of non sharp corners detected by the default values of the toolbox, a peakthreshold was used and set to 8. Below are the results obtained from the tool box.

Total Number of matches = 365

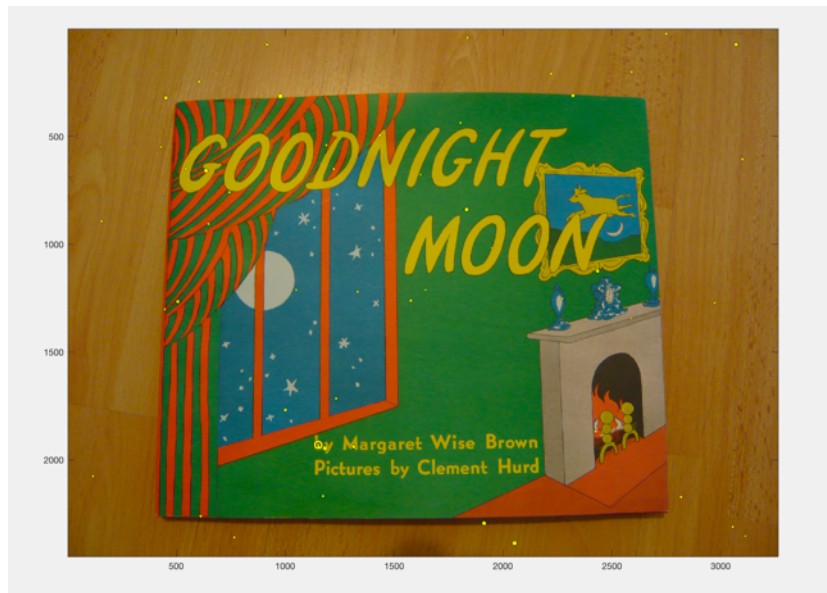


Figure 5: ToolBox Result 1

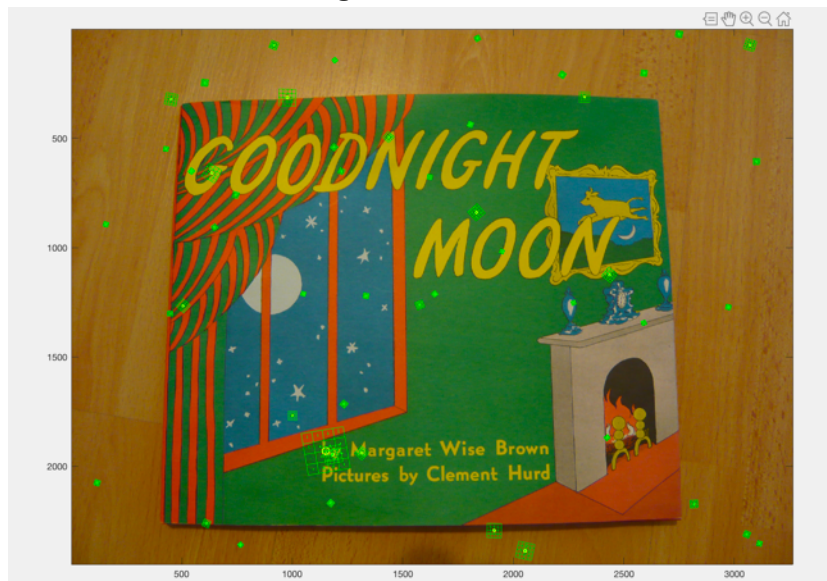


Figure 6: ToolBox Result 2



Figure 7: ToolBox Result 3

6 Conclusion

It's obvious from the results above that the SIFT feature extractor produced more accurate results than the Harris corner detector; which was expected. Also, the keypoints detected with SIFT are less noisy. The harris detector can do a better job than the above image with tweaking both thresholds (the key points threshold and the matching SSD threshold). Tweaking the thresholds was a manual effort, I am sure if I experimented with more values or an automatic tuning way; I will obtain better results for the Harris detector compared to the above results.