## Assignment 03: Feature Matching

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Part 0: Input Images

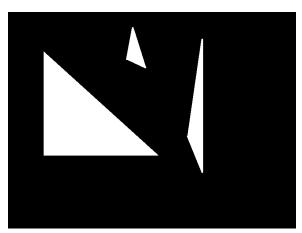


image1.png



pic1.png



pic2.png

Part 1: Synthetic Images

pic\_1\_a.png

This image was created by applying the above transformation matrix to image1.png using the cv2.warpPerspective function.



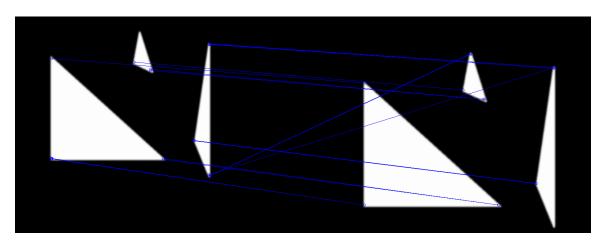




pic\_1\_b\_1.png pic\_1\_b\_2.png

These pictures were created by finding the key points of image1 and image2.

```
orb=cv2.ORB()
kp,des=orb.detectAndCompute(img,None)
img=cv2.drawKeypoints(img,kp,None,color=(0,255,0),flags=0)
```



pic\_1\_e.png

The above image was created by finding the key points of each image and matching them using the knnMatcher to pair the matches. This is the process of the given code. Then using a drawMatches functions similar to the built in cv2 function, lines between the top matches of the keypoints were drawn.

The draw part of the drawMatches function, retrieved from StackOverflow is shown below.

```
for mat in matches:
    # Get the matching keypoints for each of the images
    img1_idx = mat.queryIdx
    img2_idx = mat.trainIdx

    (x1,y1) = kp1[img1_idx].pt
```

```
(x2,y2) = kp2[img2_idx].pt
cv2.circle(out, (int(x1),int(y1)), 4, (255, 0, 0), 1)
cv2.circle(out, (int(x2)+cols1,int(y2)), 4, (255, 0, 0), 1)
cv2.line(out, (int(x1),int(y1)), (int(x2)+cols1,int(y2)),
(255, 0, 0), 1)
```

In that same function the transformation matrix was calculated form the keypoints of the two images. The order of magnitude of the matricies are the same, and the only more than trivial differences are the 1.22 vs the 1.2 of the scale factor and the 44's instead of 50's in the translate.

This pictures is image1, or the original synthetic image, transformed using the calculated transformation matrix. Ideally, this would be identical to image2 (pic\_1\_a). This was done using the warpPerspective function with the calculated transform matrix.

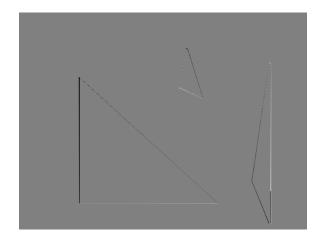
```
h2,w2=pic2.shape[:2]
pic_2_d=cv2.warpPerspective(pic1,M2,
(w2,h2))
```



pic\_1\_e.png

This picture is the difference between pic\_1\_e and image2 after normalization. If the transformation matrix was perfect, this picture would be completely gray, but the slight inconsistencies between the two pictures created the outline.

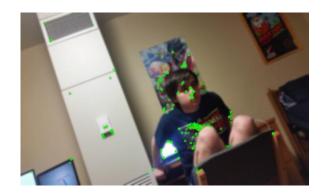
```
pic_2_e=np.float32
(funcs.grayscale(pic2))-
funcs.grayscale(pic_2_d)
normalize(pic_2_e)
```



pic\_1\_f.png

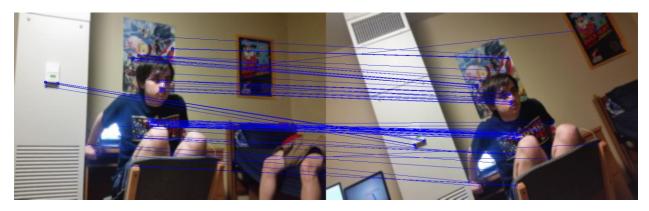
## Part 2: Real Images





pic\_2\_a\_1.png pic\_2\_a\_2.png

This is done using the same process as pic\_1\_b\_1 and pic\_1\_b\_2 after a 15-size kernel Gaussian blur.



pic\_2\_b.png

This image is created using the same process and functions as pic\_1\_c.

The matrix found from these matches is:

```
M=[[0.40479, 0.39018, 120.509], [-0.4632, 0.88074, 271.829], [-0.0001, 0.00006, 1.00000]]
```

This image is pic1 after it is translated using the above transformation matrix using the warpPerspective function. This should be as close to pic2 as possible.

```
h2,w2=pic2.shape[:2]
pic_2_d=cv2.warpPerspective(pic1,M2,
  (w2,h2))
```



pic\_2\_d.png

This image is the difference between pic\_2\_d and pic2 after normalization. This should be a gray image in the area where there are two pictures.

```
pic_2_e=np.float32(funcs.grayscale
(pic2))-funcs.grayscale(pic_2_d)
normalize(pic_2_e)
```



pic\_2\_e.png

This image is pic2 and the non-clipped version of pic\_2\_d overlaid on top of each other.

```
def getBoundingRectangle(w,h,M):
    For corner:
        p1 =
    M.dot(np.float32([[x], [y],
        [1]]))
    return p1[0, 0] / p1[2, 0],
        p2[1, 0] / p2[2, 0],
        p3[0, 0] / p3[2, 0],
        p4[1, 0] / p4[2, 0]
```

