

# Homework 2

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## Part 1

### Task 1.5

Image (model\_chickenbroth.jpg) with the detected keypoints:



## Part 2

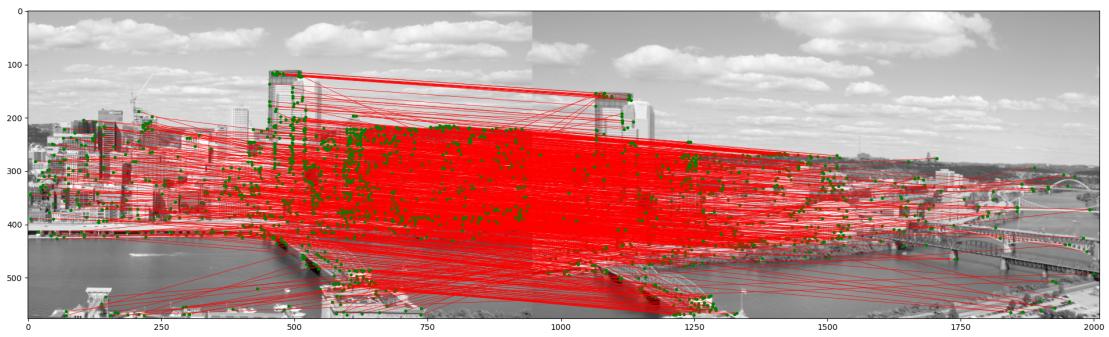
### Task 2.4

Results of BRIEF descriptor matching for 3 sets of images:

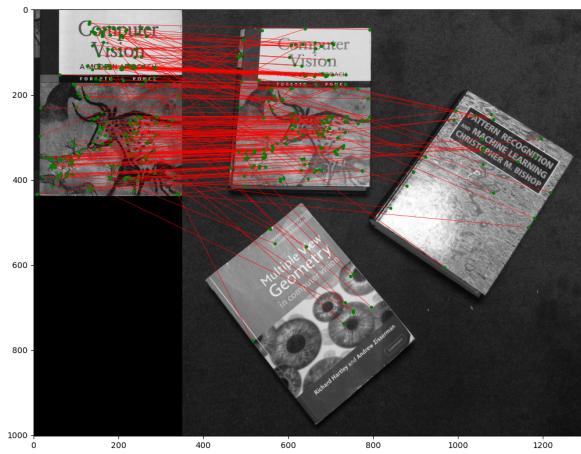
1. Left: model\_chickenbroth.jpg, right: chickenbroth\_01.jpg



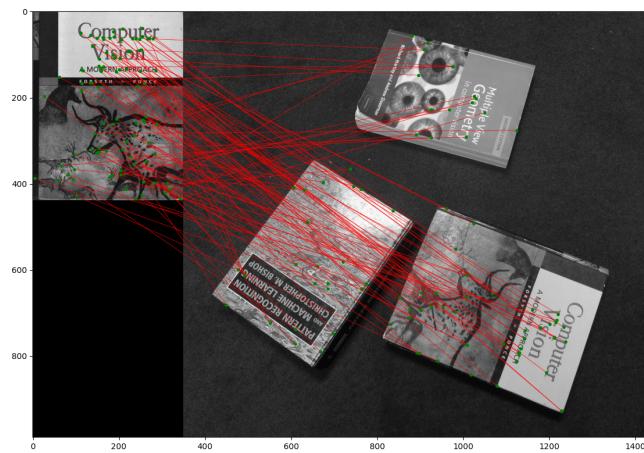
2. Left: incline\_L.png, right: incline\_R.png



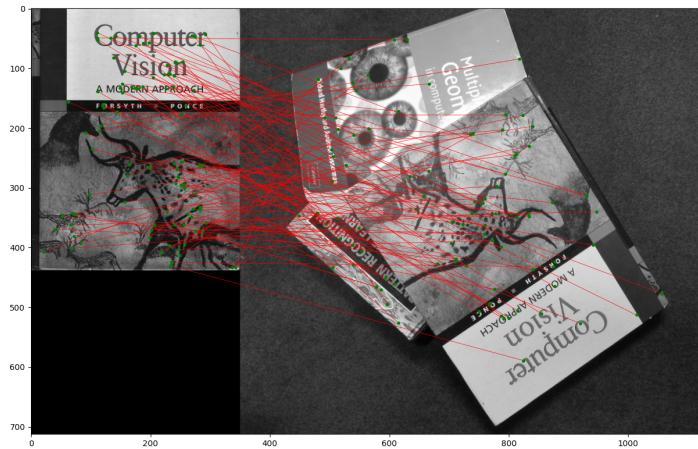
3. Left: pf\_scan\_scaled.jpg, right: pf\_floor.jpg



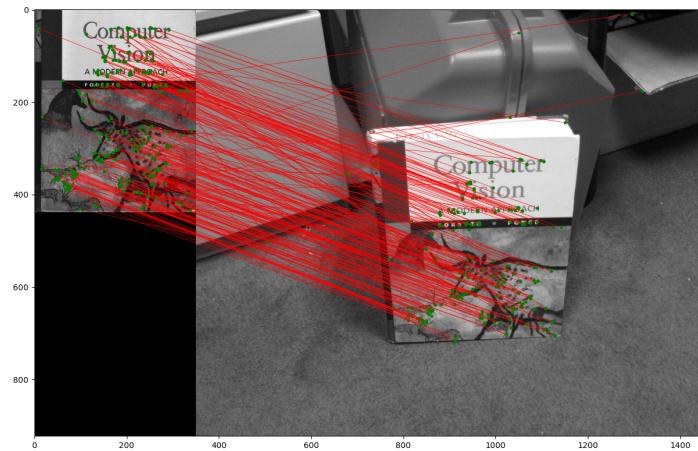
4. Left: pf\_scan\_scaled.jpg, right: pf\_floor\_rot.jpg



5. Left: pf\_scan\_scaled.jpg, right: pf\_pile.jpg



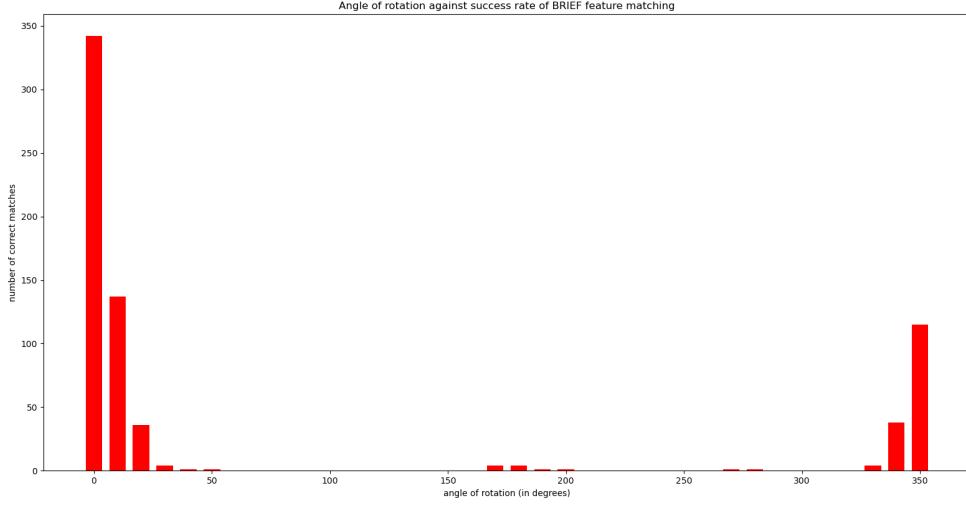
6. Left: pf\_scan\_scaled.jpg, right: pf\_stand.jpg



**Discussion:** The pf\_pile.png and pf\_floor\_rot.png image have the worst performance, with over half of the features wrongly mapped to other books and having much less matches than other images. The pf\_scan\_scaled.jpg also has incorrectly matched features. In contrast, the other images perform rather well, with most keypoints correctly mapped.

## Task 2.5

Required bar graph:



The algorithm has relative high accuracy when the image is only rotated slightly, and completely fails at rotation angles greater than 20 degrees.

I believe this is due to the way the test point pairs are selected in BRIEF. Since the tests will compare the intensities between corresponding pixels in a patch around a keypoint, when the image is rotated by a large angle, the pixels being compared does not have a correspondence relationship. Slight rotations are counteracted by gaussian blur which merges rotated pixels, but large rotation will completely move the pixel away. That is why BRIEF descriptors are not rotation-invariant.

## Part 3

### Part (a)

For a point  $w$  on the common plane,  $\tilde{u}$  and  $\tilde{x}$  are the 2D projected point on the first and second camera respectively. The two are related by a relationship

$$\lambda \tilde{x} = H \tilde{u}$$

For  $N$  pairs of such points, we have

$$\begin{aligned} \lambda_n \widetilde{x_n} &= H \widetilde{u_n} \\ \lambda_n \begin{bmatrix} \widetilde{x_{xn}} \\ \widetilde{x_{yn}} \\ 1 \end{bmatrix} &= \begin{bmatrix} H_{11} & H_{12} & H_{13} \\ H_{21} & H_{22} & H_{23} \\ H_{31} & H_{32} & H_{33} \end{bmatrix} \begin{bmatrix} \widetilde{u_{xn}} \\ \widetilde{u_{yn}} \\ 1 \end{bmatrix} = \begin{bmatrix} -h_1^T \\ -h_2^T \\ -h_3^T \end{bmatrix} \widetilde{u_n} \end{aligned}$$

We let  $h = [ H_{11} \quad H_{12} \quad \dots \quad H_{33} ]^T$ . Then for one set of points,

$$\begin{aligned} \lambda_n \widetilde{x_{xn}} &= h_1^T \widetilde{u_n} \\ \lambda_n \widetilde{x_{yn}} &= h_2^T \widetilde{u_n} \\ \lambda_n &= h_3^T \widetilde{u_n} \end{aligned}$$

which can be reduced to

$$\begin{aligned} h_1^T \widetilde{u_n} - (h_3^T \widetilde{u_n}) \widetilde{x_{xn}} &= 0 \\ h_2^T \widetilde{u_n} - (h_3^T \widetilde{u_n}) \widetilde{x_{yn}} &= 0 \end{aligned}$$

In matrix form,

$$A_n = \begin{bmatrix} \widetilde{u_n}^T & 0 & -\widetilde{x_{xn}}\widetilde{u_n}^T \\ 0 & \widetilde{u_n}^T & -\widetilde{x_{yn}}\widetilde{u_n}^T \end{bmatrix} \begin{bmatrix} h_1 \\ h_2 \\ h_3 \end{bmatrix}$$

Repeat this  $N$  times to get

$$A = \begin{bmatrix} A_1 \\ A_2 \\ \vdots \\ A_N \end{bmatrix}$$

### Part (b)

There are 9 elements in  $h$ .

### Part (c)

$H$  has 9 elements, but is only defined up to scale (because it relates homogenous coordinates). Hence, it only has  $9 - 1 = 8$  degrees of freedom.

Each point correspondence gives two linear equations (derived above).

Therefore, we need  $8/2 = 4$  point correspondences to solve the system.

### Part (d)

1. Compute the SVD of  $A$  to get  $U\Sigma V^T$ .

2. The solutions for  $h$  is given by the rightmost column of the matrix  $V$ .

## Part 6

### Task 6.3

This is the final unclipped panorama:



## Part 7

The code for this section can be found in `augmentedReality.py`.

## Task 7.2

Here is the final image:

