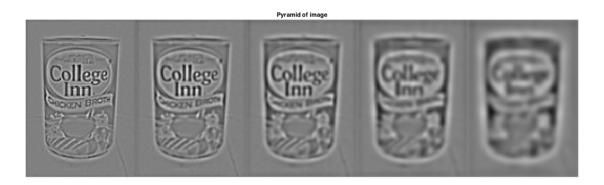
Jerry Hsiung Computer Vision 16-720 Assignment 2 October 9, 2016

Assignment #2

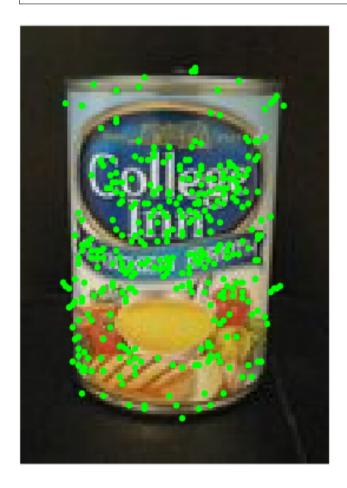
1.1 Gaussian Pyramid of model_chickenbroth.jpg



1.2 Diference in Guassian Pyramid of *model_chickenbroth.jpg*



1.5 Image of detected keypoints with edge suppresions on *model_chickenbroth.jpg*



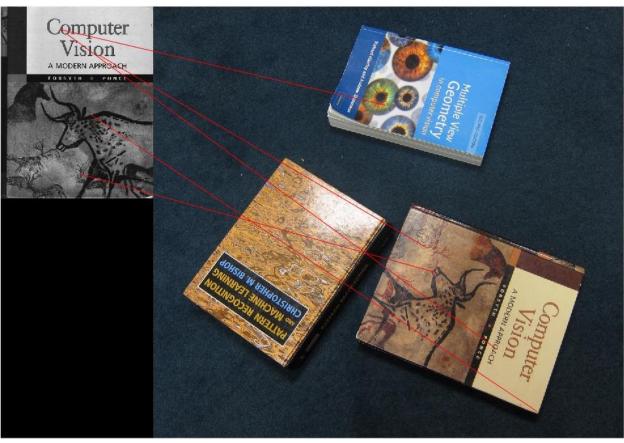
2.4 Results of two images matching on *chickenbroth.jpg*, *inline.jpg*, and *textbook.jpg*.

The matching results are below (Running all the matching with ratio = 0.7)





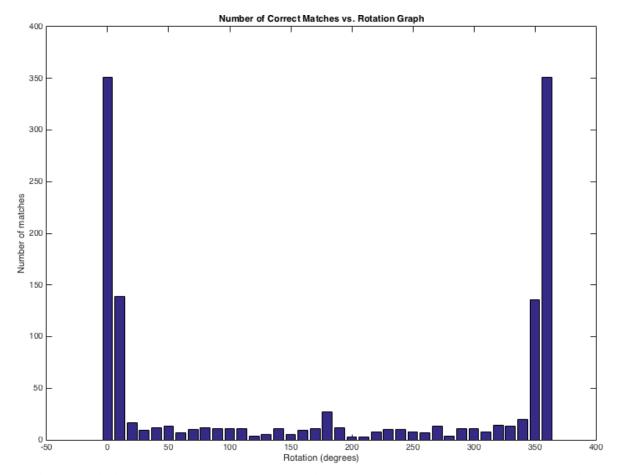




As we can see from above, after rotation, the book is not matched well. This is because BRIEF is not rotation-invariant (extra credit). This is because the tests pairs that we cre-

ated do not calculate the same histogram after the image is rotated. One possible way to fix this is to rotate the test pairs too so the local environment in a descriptor will be consistent around all the interest points.

2.5 Show the histogram of performance of matching with different rotation of *model_chickenbroth.jpg*.



As discussed above (2.4), the main reason that BRIEF is not rotation invariant is because the testpairs don't allow correct histogram calculation upon rotation.

3.1

- (a) Write out the expression of A, in homography Ah = 0.
- (b) How many elements are in h?
- (c) How many point pairs (correspondences) are required to solve this system? *Hint:* How many degrees of freedom are in **h**? How much information does each point correspondence give?
- (d) Show how to estimate the elements in **h** to find a solution to minimize this homogeneous linear least squares system. Step us through this procedure. *Hint:* Use the Rayleigh quotient theorem (homogeneous least squares).
- (a) Given that $p^i \equiv \mathbf{H}q^i$, where $p^i = [x'_i, y'_i, 1]$, $q^i = [x_i, y_i, 1]$, we can expand it out and get the equation as:

$$\begin{pmatrix} x_i' \\ y_i' \\ 1 \end{pmatrix} = \begin{pmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{pmatrix} \begin{pmatrix} x_i \\ y_i \\ 1 \end{pmatrix}$$

so divide x'_i and y'_i with the third row, we get:

$$\frac{x_i'}{1} = \frac{h_{11}x_i + h_{12}y_i + h_{13}}{h_{31}x_i + h_{32}y_i + h_{33}}$$
$$\frac{y_i'}{1} = \frac{h_{21}x_i + h_{22}y_i + h_{23}}{h_{31}x_i + h_{32}y_i + h_{33}}$$

Rearrange, we get:

$$x_i'(h_{31}x_i + h_{32}y_i + h_{33}) = h_{11}x_i + h_{12}y_i + h_{13}$$

$$h_{31}x_ix_i' + h_{32}y_ix_i' + h_{33}x_i' - h_{11}x_i + h_{12}y_i + h_{13} = 0$$

$$y_i'(h_{31}x_i + h_{32}y_i + h_{33}) = h_{21}x_i + h_{22}y_i + h_{23}$$

$$h_{31}x_iy_i' + h_{32}y_iy_i' + h_{33}y_i' - h_{21}x_i + h_{22}y_i + h_{23} = 0$$

Now put it in the matrix $\mathbf{Ah} = 0$, we have the desired homogeneous system for all N points:

$$\begin{pmatrix} x_1 & y_1 & 1 & 0 & 0 & 0 & -x_1x'_1 & -y_1x'_1 & -x'_1 \\ 0 & 0 & 0 & x_1 & y_1 & 1 & -x_1y'_1 & -y_1y'_1 & -y'_1 \\ \vdots & \vdots \\ x_N & y_N & 1 & 0 & 0 & 0 & -x_Nx'_N & -y_Nx'_N & -x'_N \\ 0 & 0 & 0 & x_N & y_N & 1 & -x_Ny'_N & -y_Ny'_N & -y'_N \end{pmatrix} \begin{pmatrix} h_{11} \\ h_{12} \\ h_{13} \\ h_{21} \\ h_{22} \\ h_{23} \\ h_{31} \\ h_{32} \\ h_{33} \end{pmatrix}$$

Now, if we have *N* point, then the matrix *A* will have $2N \times 9$ dimension.

- (b) From above, we can see that **h** is 9×1 , which has 9 elements.
- (c) **h** has 9 unknown, however, it is scale-invariant, which means that we can multiply by a constant that eliminates one of the degree of the freedom. So **h** has 8 degress of freedom. Since from part (a) we see that each point provides two equations, we need a total of 4 non-colinear point pairs to solve the unique system.
- (d) We are trying to find \mathbf{h} such that it minimizes \mathbf{Ah} . Using the given hint and Rayleigh's quotient theorem for solving homogeneous least squares (let $\mathbf{h} = x$). First we set up our problem as:

$$argmin_h \frac{||Ax||^2}{||x||^2}$$
, where $||x||^2 = 1$

Rearrange, we get:

$$\frac{||Ax||^2}{||x||^2} = \frac{x^T A^T A x}{x^T x}$$

which is in Rayleigh's quotient's form. We now express A^TA with its eigenvalues and eigenvectors as:

$$A^T A v_i = \lambda_i v_i$$

, we can also express x as the sum of on the same basis of eigenvectors v_i ,

$$x = \sum \alpha_i v_i$$

So we can rewrite above as:

$$\frac{||Ax||^2}{||x||^2} = \frac{x^T A^T A x}{x^T x}$$

$$= \frac{(\sum \alpha_i v_i)^T A^T A (\sum \alpha_i v_i)}{(\sum \alpha_i v_i)^T (\sum \alpha_i v_i)}$$

$$= \frac{\sum \alpha_i^2 \lambda_i}{\sum \alpha_i^2}$$

Note we can simplified $v_i v_j = 0$ for $i \neq j$ because eigenvectors are orthogonal. To minimize this, we can cearly see that we want to choose the smallest eigenvalues λ_{min} in A^TA . This is to choose **h** to be the the corresponding eigenvector in A^TA . Thus we have shown that **h** is the eigenvector corresponding to the smallest eigenvalues in A. (In SVD decomposition of $A = U\Sigma V^T$, this is a column of V that corresponds to the smallest eighenvalues in Σ .)

5.1 Show the warped *incline_R.png*.

below is the wraped *incline_R.png*:



also the stiched image:



5.2 Show the warped *incline_R.png* with no image clipping.



6.2 Show the final panorama image with image mask blending:

The default iteration and tolerance for RANSAC are 2000 and 5:



9