
Part 1.

- (1) We believe that the grouping algorithm graph-cuts, would be the most appropriate grouping algorithm to recover the model parameter hypotheses from the continuous vote space. The reason is that it links every pair of pixels and assigns an affinity weight for each edge.
Also, graph-cuts connect shapes with edges that have high affinity while mean shift and k-means does not, which suits those lines and circles defined by a set of boundary points in Hough Transform.
- (2) From using K-means clustering with two groups, we will have two center points which divides all the feature inputs into two clusters. Each center point will get half of the feature points and the center point will be positioned by calculating the least squared Euclidean distance. After that it will update the new center of each cluster. The process will keep repeated until half of the data points are in one group and the other half is in the other group.
- (3) TO BE CONTINUE.

Part 2.

- (a) The code is written in file **get_correspondences.m**. The function has two inputs of two images and the number of points the user wants to pick. The outputs are two $2 \times N$ point matrices of selected points.
- (b) The code is written in file **computeH.m**.
- (c) The code is written in file **warpImage.m**.
- (d) Images:

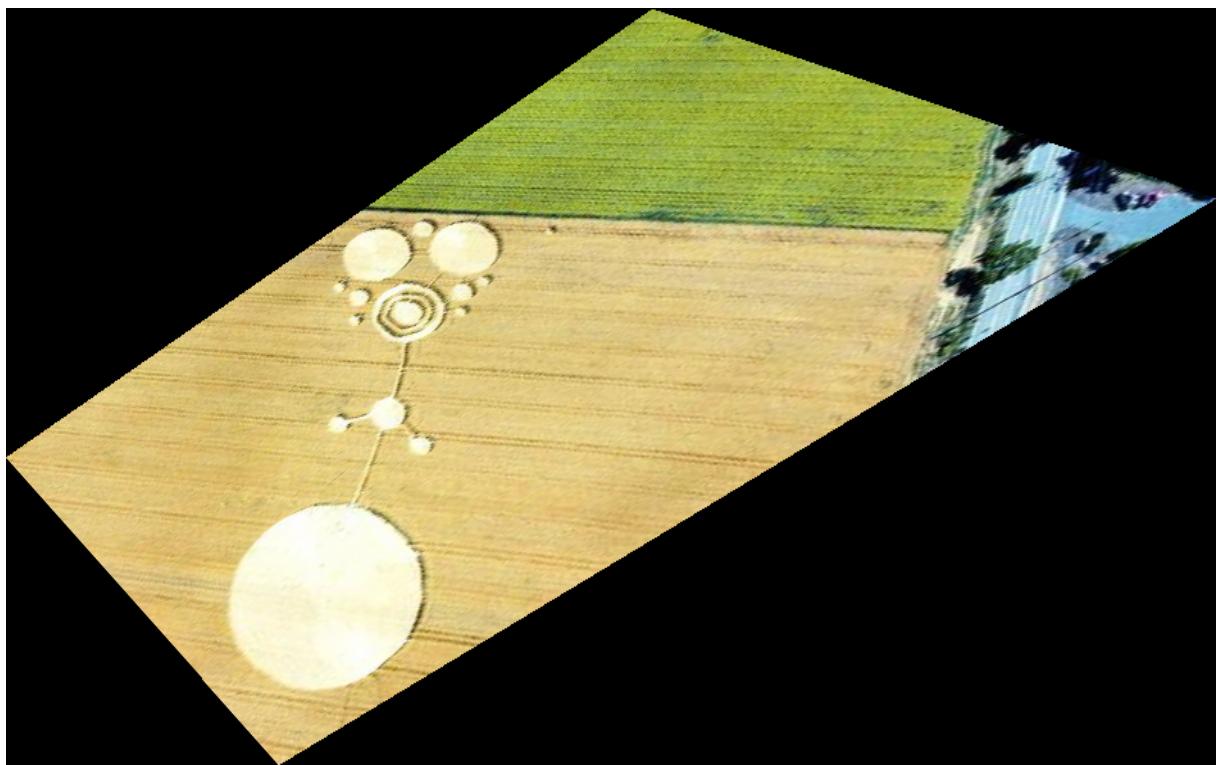


Figure 1: crop_warp

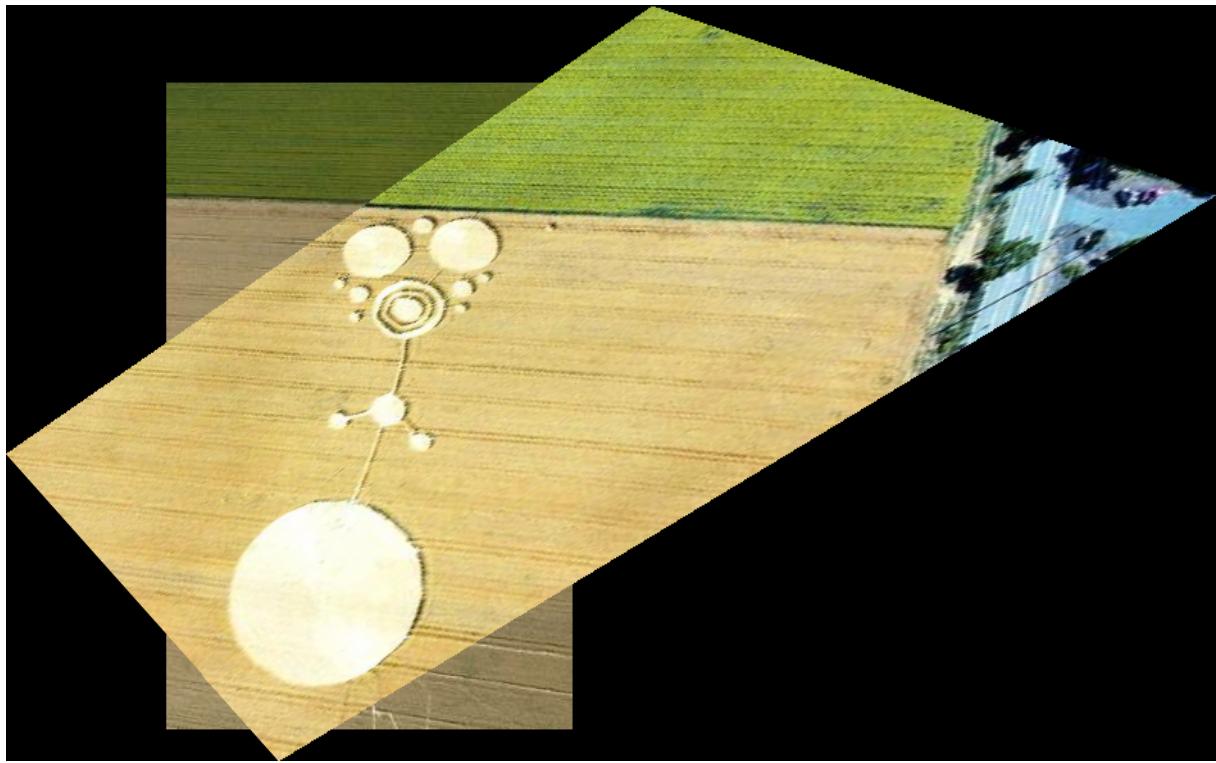


Figure 2: crop_merge

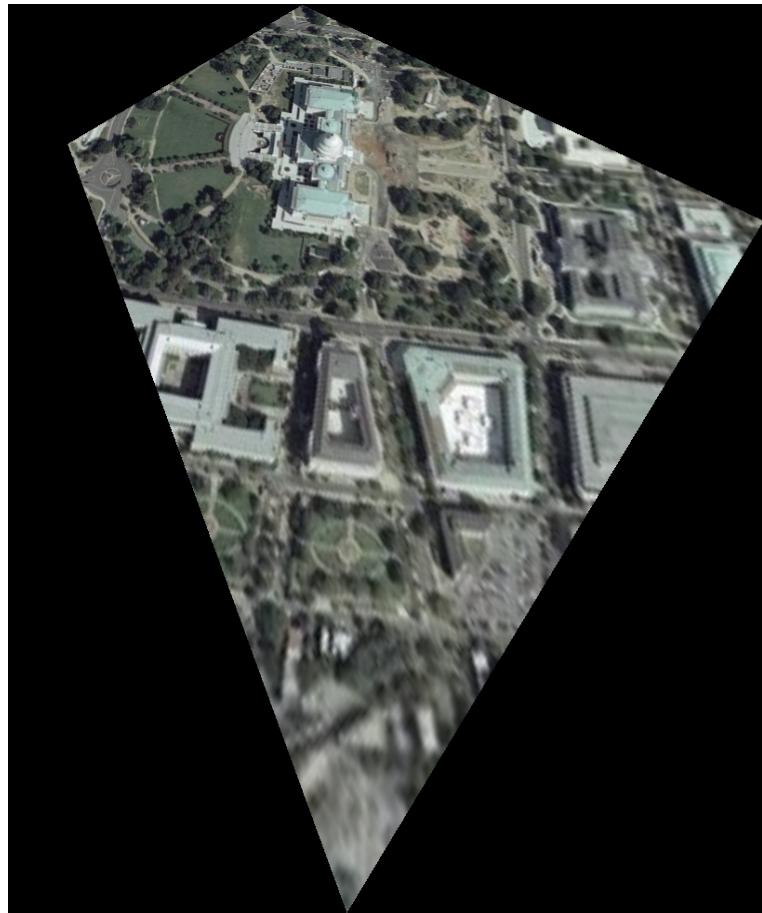


Figure 3: wdc_warp

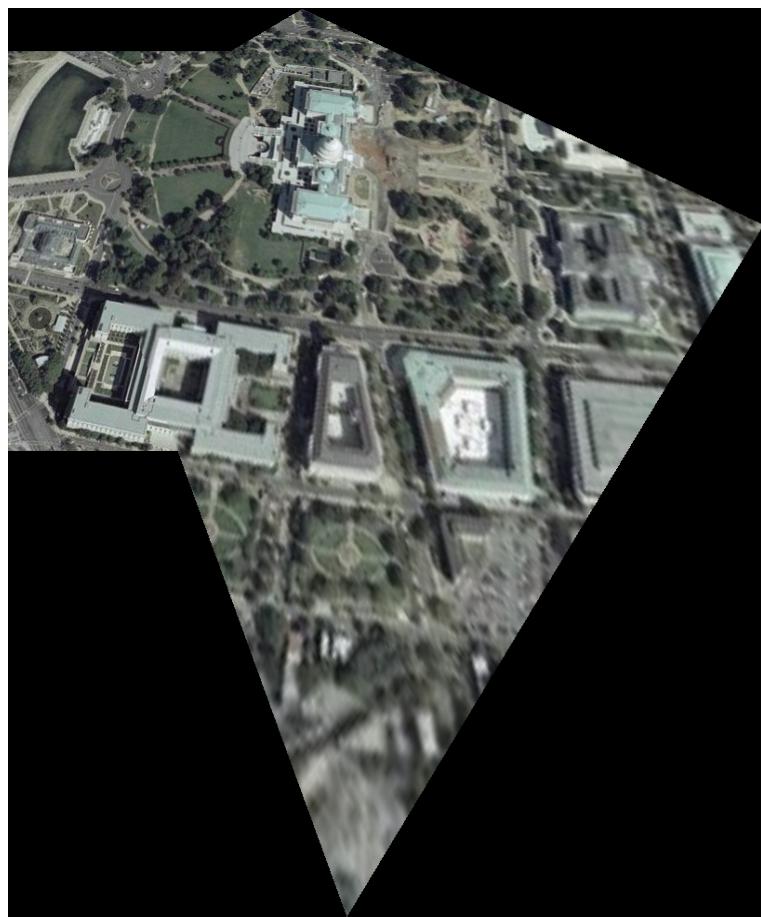


Figure 4: wdc_merge

The data points1 and points2 are saved in the file **points.m**.

(e) Images:



Figure 5: town_source_1



Figure 6: town_source_2



Figure 7: town_mosaic

Source: <https://www.bilibili.com/bangumi/play/ep264278>.

Another example:



Figure 8: mountain_source_1



Figure 9: mountain_source_2



Figure 10: mountain_mosaic

Source: <https://unsplash.com/photos/Y8lCoTRgHPE>.

(f) Images:



Figure 11: plate



Figure 12: car



Figure 13: car_mosaic

Source: ScreenShots from Game Forza Horizon4.

Take the license plate picture, and mosic it on the car. □

Part 3.

(a) Images:



Figure 14: mountain_source_1



Figure 15: mountain_source_2



Figure 16: mountain_select_1



Figure 17: mountain_select_2



Figure 18: mountain_mosaic_original



Figure 19: mountain_mosaic_RANSAC

Source: <https://unsplash.com/photos/Y8lCoTRgHPE>.

Code: **RANSAC_script.m** and **RANSAC.m**.

Explanation: We implemented a new function **bestH = RANSAC(t1, t2)** to calculate the new H value using RANSAC algorithm. From Figure 16 and Figure 17, we can see there is one bad point selected. Thus, in Figure 18, the image is "broken" and no longer seems "stitch". In Figure 19, we use RANSAC method, and that can eliminate the bad point; therefore, gives us the correct stitch.

Implementation: The user can use the script **RANSAC_script.m** and change the image names to get results with RANSAC algorithm.

(b) Images:

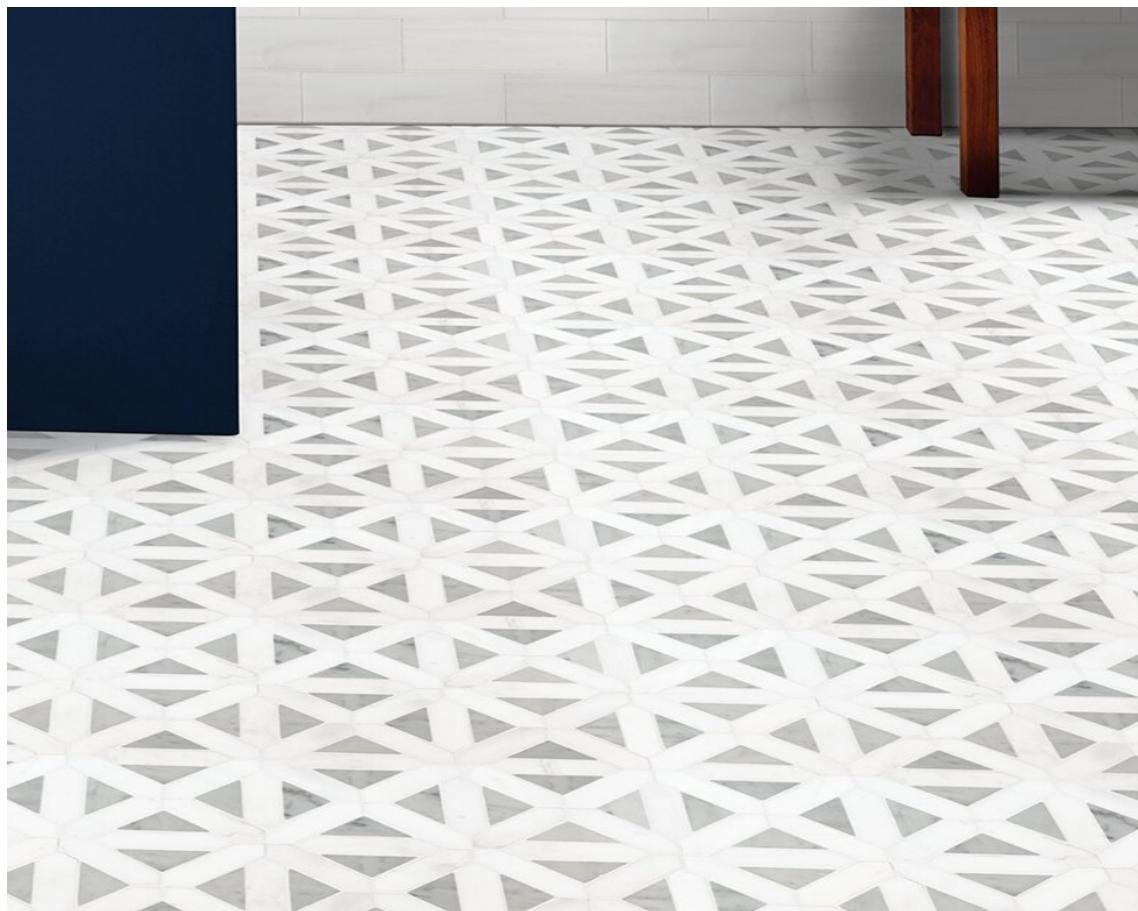


Figure 20: tiles_source



Figure 21: tiles_fronto

Source: <https://www.wayfair.com/home-improvement/pdp/msi-bianco-dolomite-marble-mosaic-tile-mvp4112.html?piid=49665325>.

Code: **fronto_script.m** and **get_correspondences_fronto.m**.

Explanation: We revised the **get_correspondences()** function to only get the four points of one image. Then, we use these four points as the first input points to compute H, and we selected the four corners of the image as the second input points to compute H. Lastly, as usual, we use the image twice and H to get the results.

Implementation: The user can use the script **fronto_script.m** and change the image name to get results. Either warp result or merge result is okay. \square