

21501886

HW3

I started the code by getting all images with a for loop. I used `slicomex` function since it looked smoother and I initially used 500 superpixels, but after some trials, I decided to use 500 superpixels since it made the end result better. After that, I went on with getting RGB values of all pixels in the superpixels. After getting values, I calculated their mean value dynamically. For this part, I only used the image as my parameter and my output was a cell that includes RGB mean values of superpixels. It has the number of superpixels as its rows and 1 as its column. In each column, there are RGB mean values. Then, I took the transpose of the cell so that I can normalize mean values easily. After taking the transpose, my cell had 3 rows (RGB) and the number of superpixels as the columns. I also convert my cells to an array and normalize the values. After normalizing the values, I calculated the colour distance for two superpixels as $D = \sqrt{((R1 - R2)^2 + (G1 - G2)^2 + (B1 - B2)^2)}$. I put my threshold as 0.11 after doing many trials. It was not optimized for all pictures but this was the one of best I can get. The output was a label array that I later create a mask with it.

After calculating the colour labels, I went on with Gabor filters. I converted my image to grayscale and create a cell with four rows and four columns. I used 2, 3.3, 4.6 and 6 as my scales. I first used 2, 4, 6 and 8 but then I realized scale 8 was very big and it blocks nearly all details, so I changed my scales. Also, I used 45, 90, 135 and 180 as my orientation. I applied `imgaborfilt` filter and get the magnitudes. In figure 1, you can see all Gabor filters for all images.

Scale : 1 Orientation : 1



Scale : 1 Orientation : 1



Scale : 1 Orientation : 3



Scale : 1 Orientation : 4



Figure 1: Bus Image for Scale 2 and Orientations 45-90-135-180

Scale : 2 Orientation : 1



Scale : 2 Orientation : 2



Scale : 2 Orientation : 3

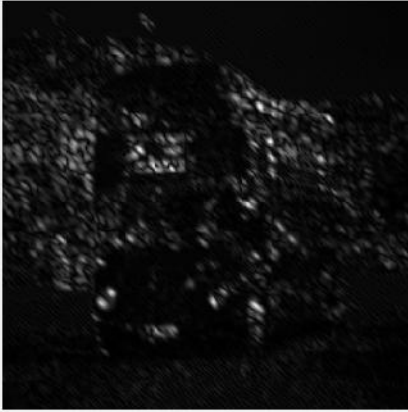


Scale : 2 Orientation : 4

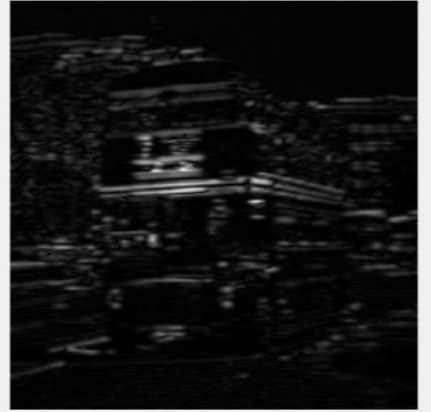


Figure 2: Bus Image for Scale 3.3 and Orientations 45-90-135-180

Scale : 3 Orientation : 1



Scale : 3 Orientation : 2



Scale : 3 Orientation : 3



Scale : 3 Orientation : 4

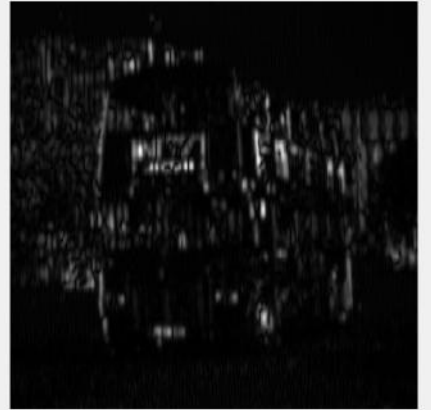
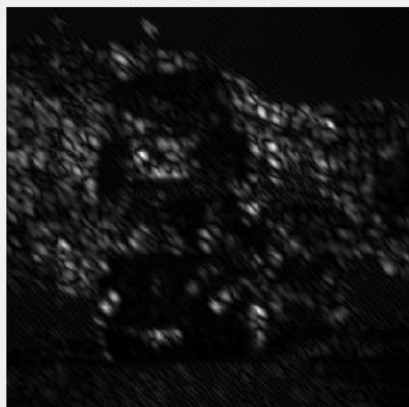
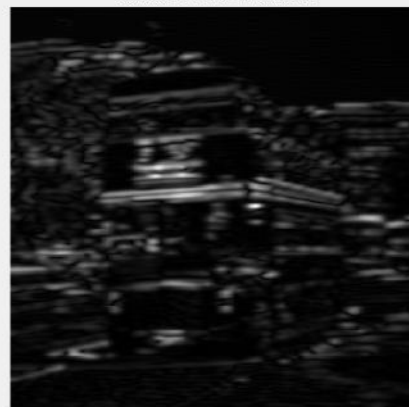


Figure 3: Bus Image for Scale 4.6 and Orientations 45-90-135-180

Scale : 4 Orientation : 1



Scale : 4 Orientation : 2



Scale : 4 Orientation : 3



Scale : 4 Orientation : 4

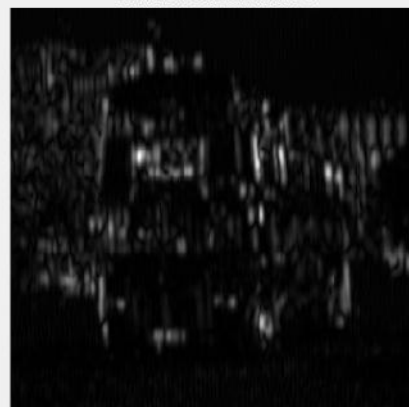


Figure 4: Bus Image for Scale 6 and Orientations 45-90-135-180

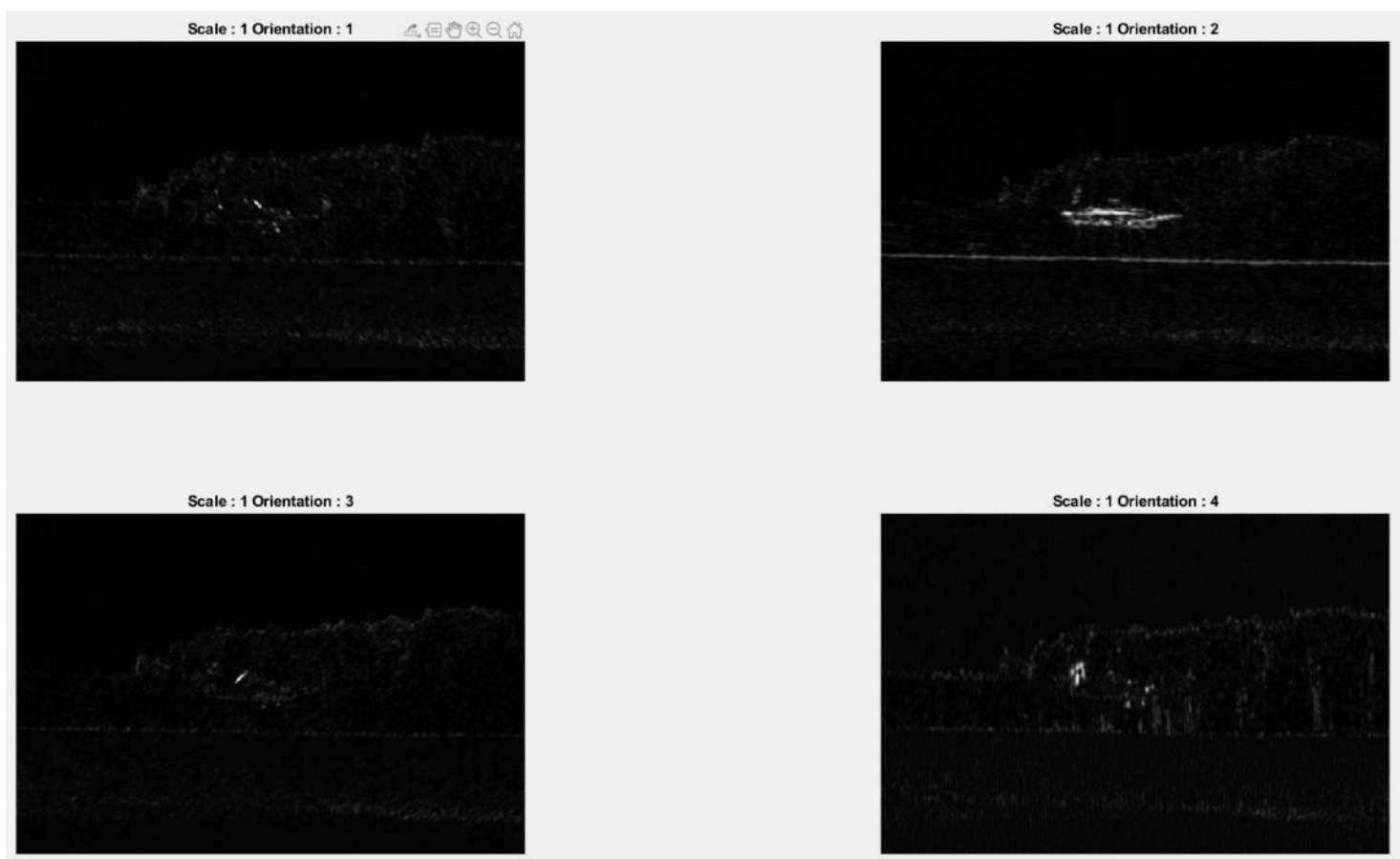


Figure 5: Plane Image for Scale 2 and Orientations 45-90-135-180

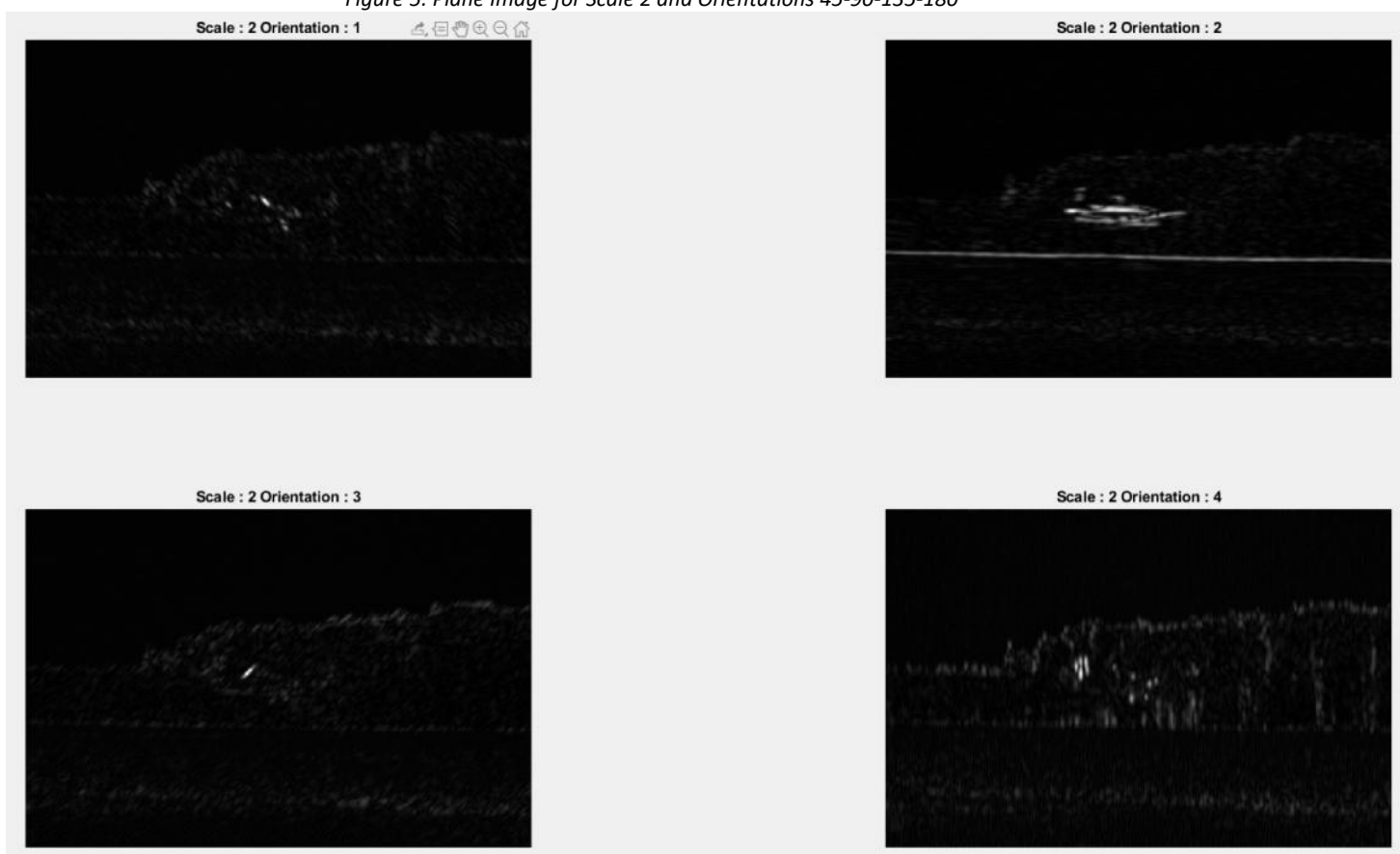


Figure 6: Plane Image for Scale 3.3 and Orientations 45-90-135-180

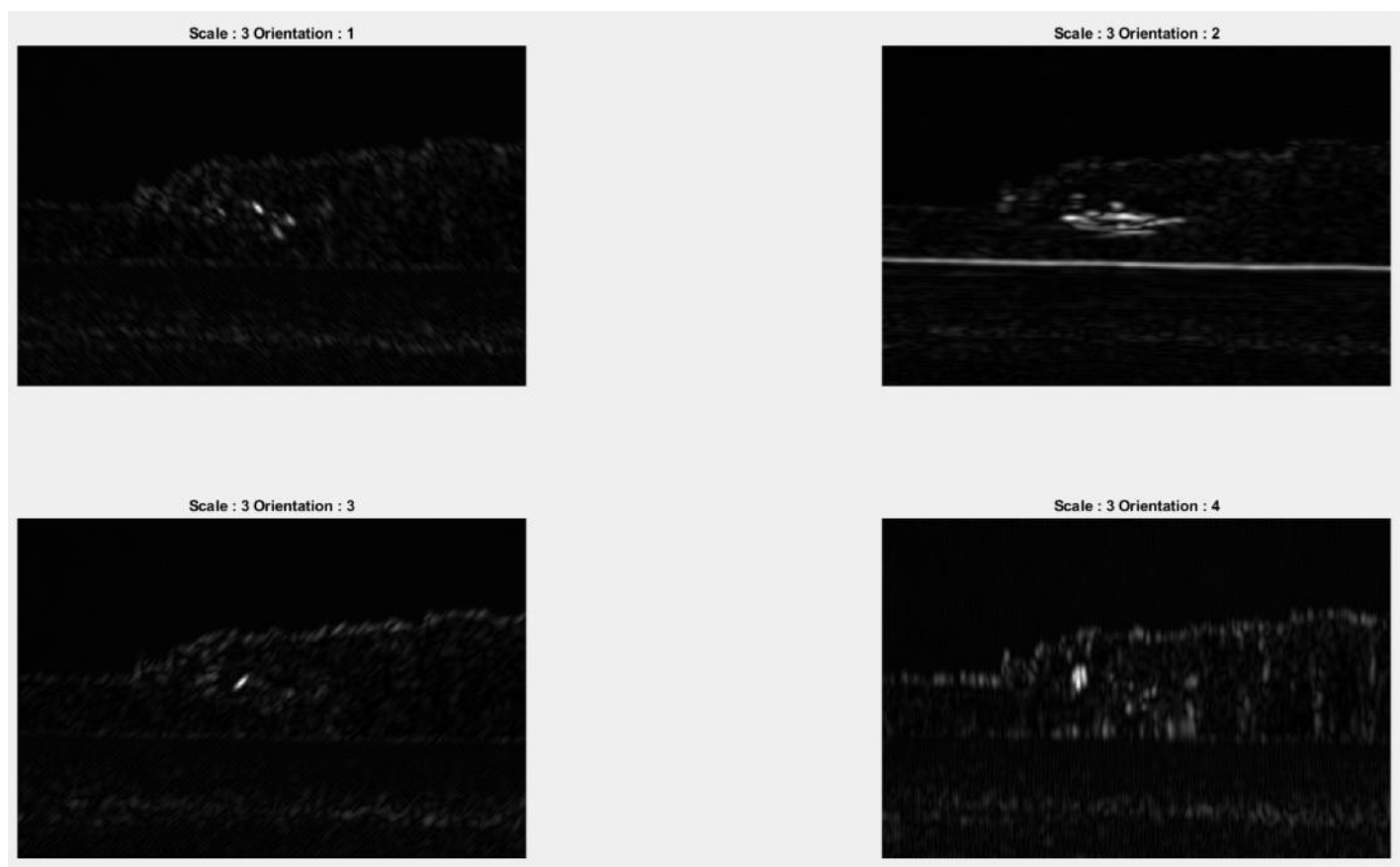


Figure 7: Plane Image for Scale 4.6 and Orientations 45-90-135-180

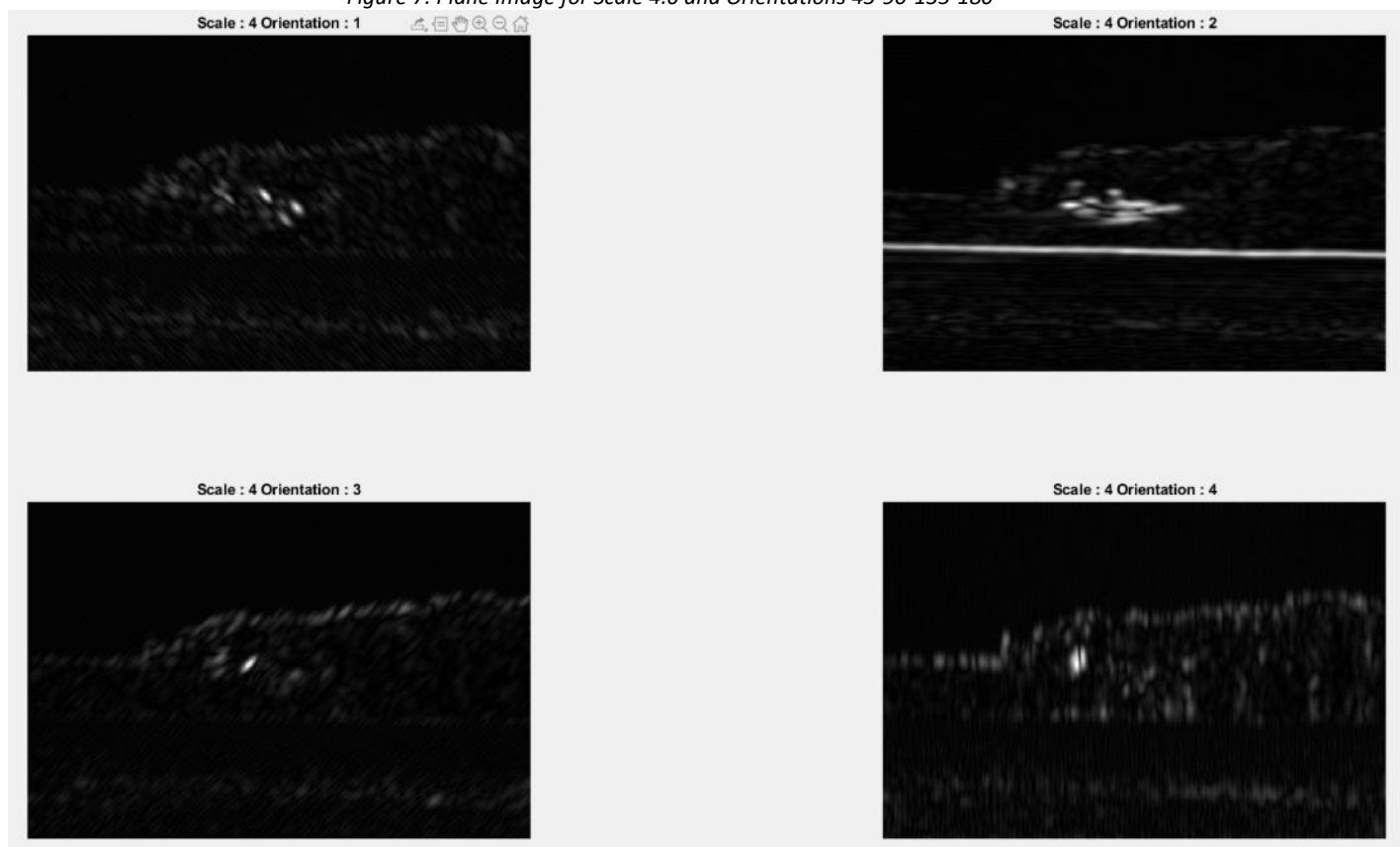


Figure 8: Plane Image for Scale 6 and Orientations 45-90-135-180

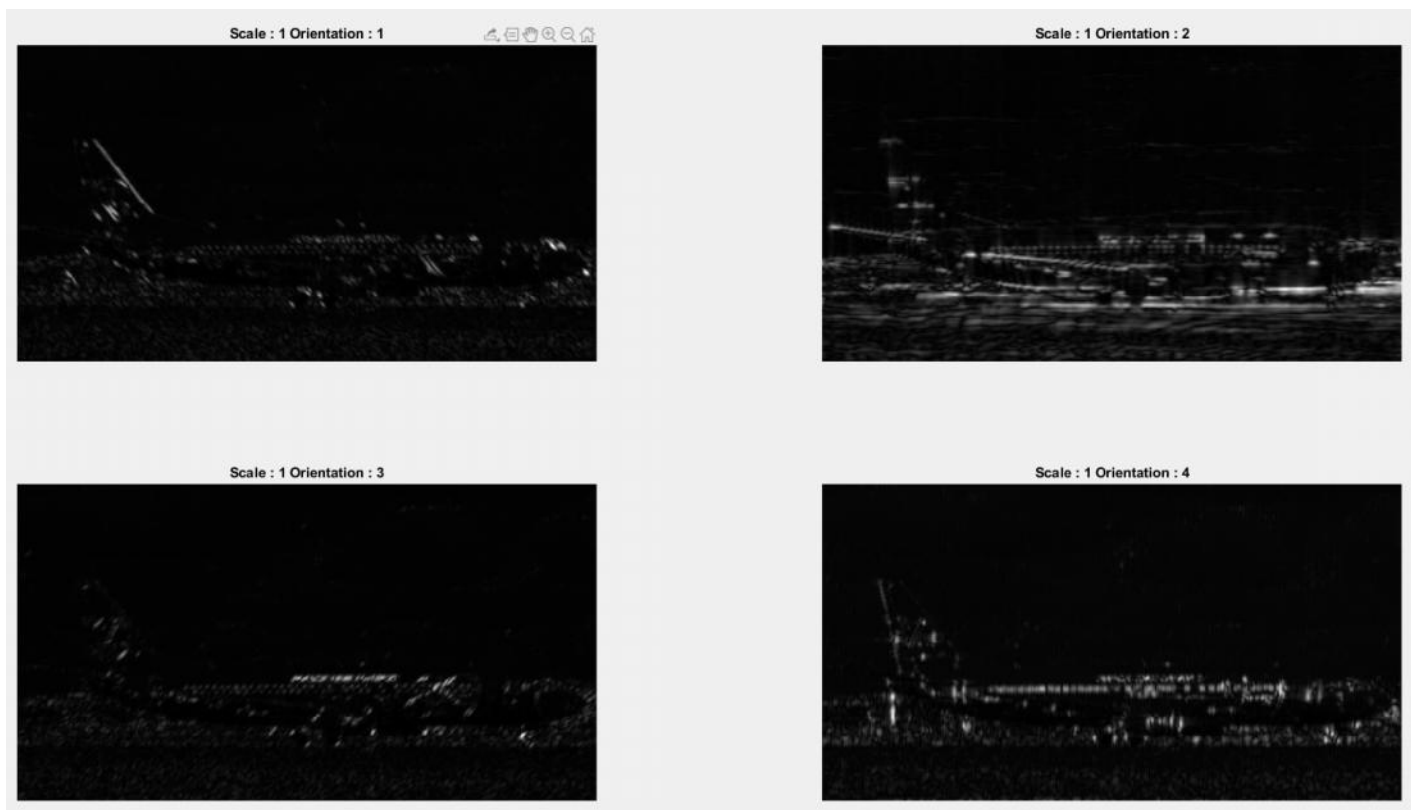


Figure 9: Plane Image for Scale 2 and Orientations 45-90-135-180

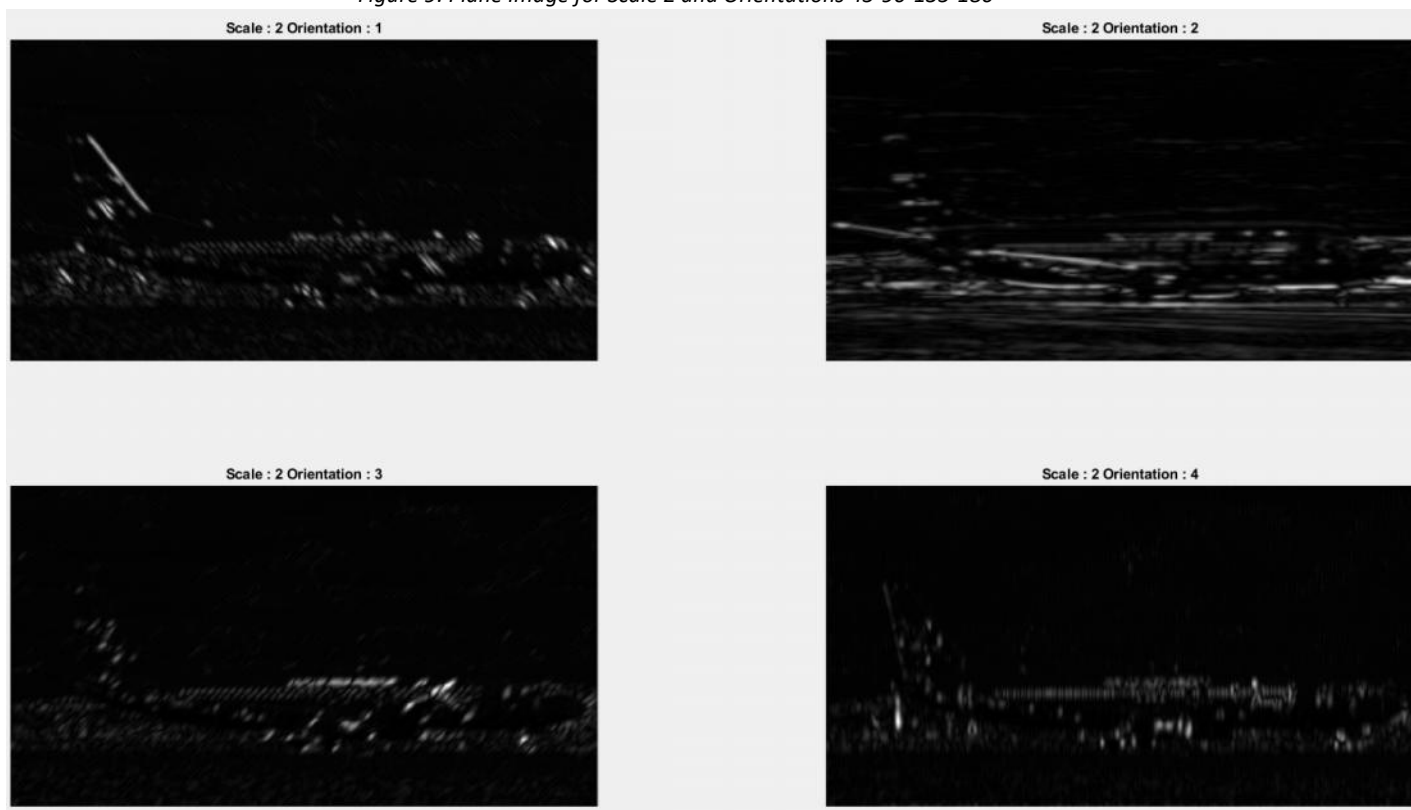


Figure 10: Plane Image for Scale 3.3 and Orientations 45-90-135-180

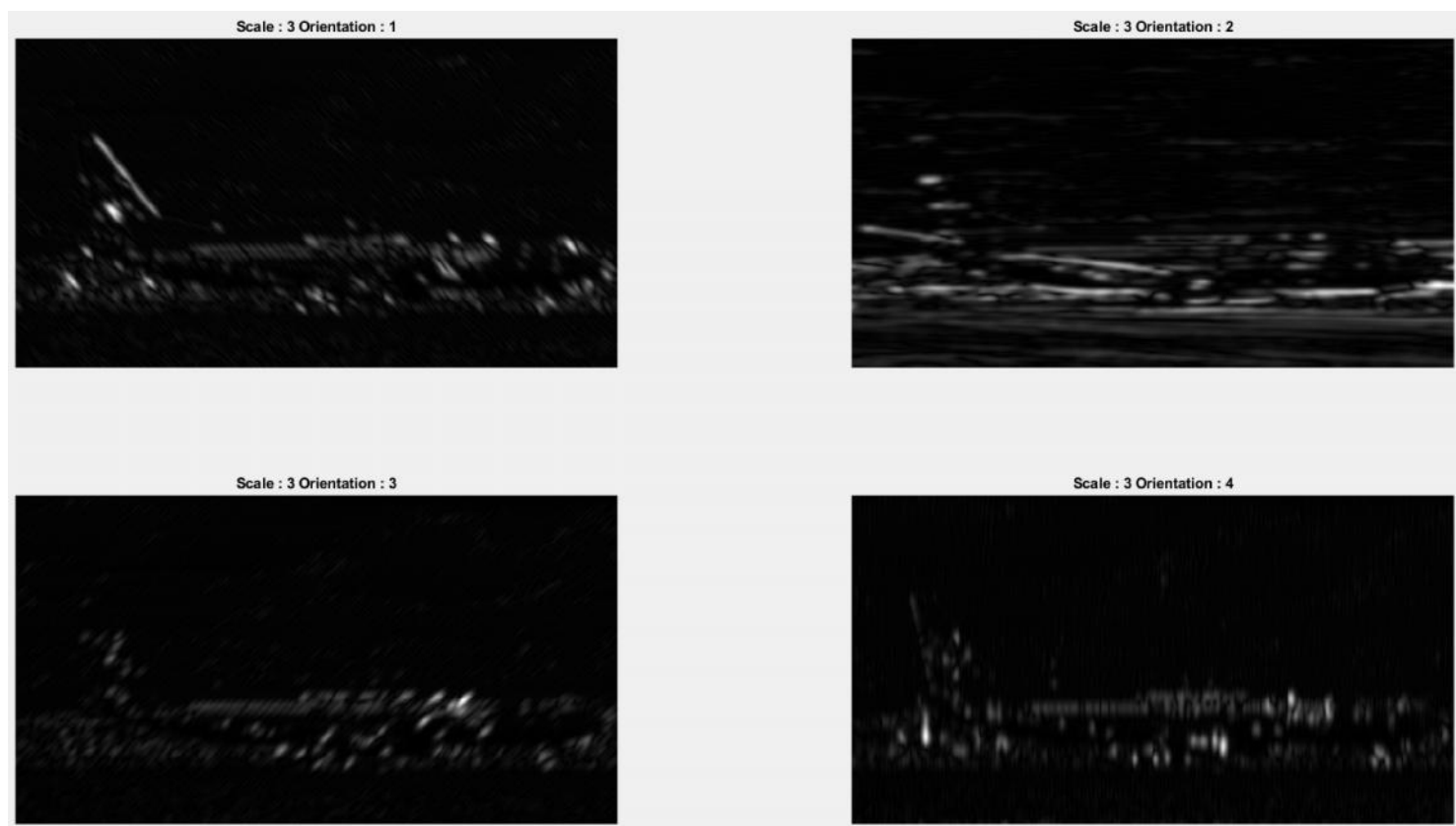


Figure 11: Plane Image for Scale 4.6 and Orientations 45-90-135-180

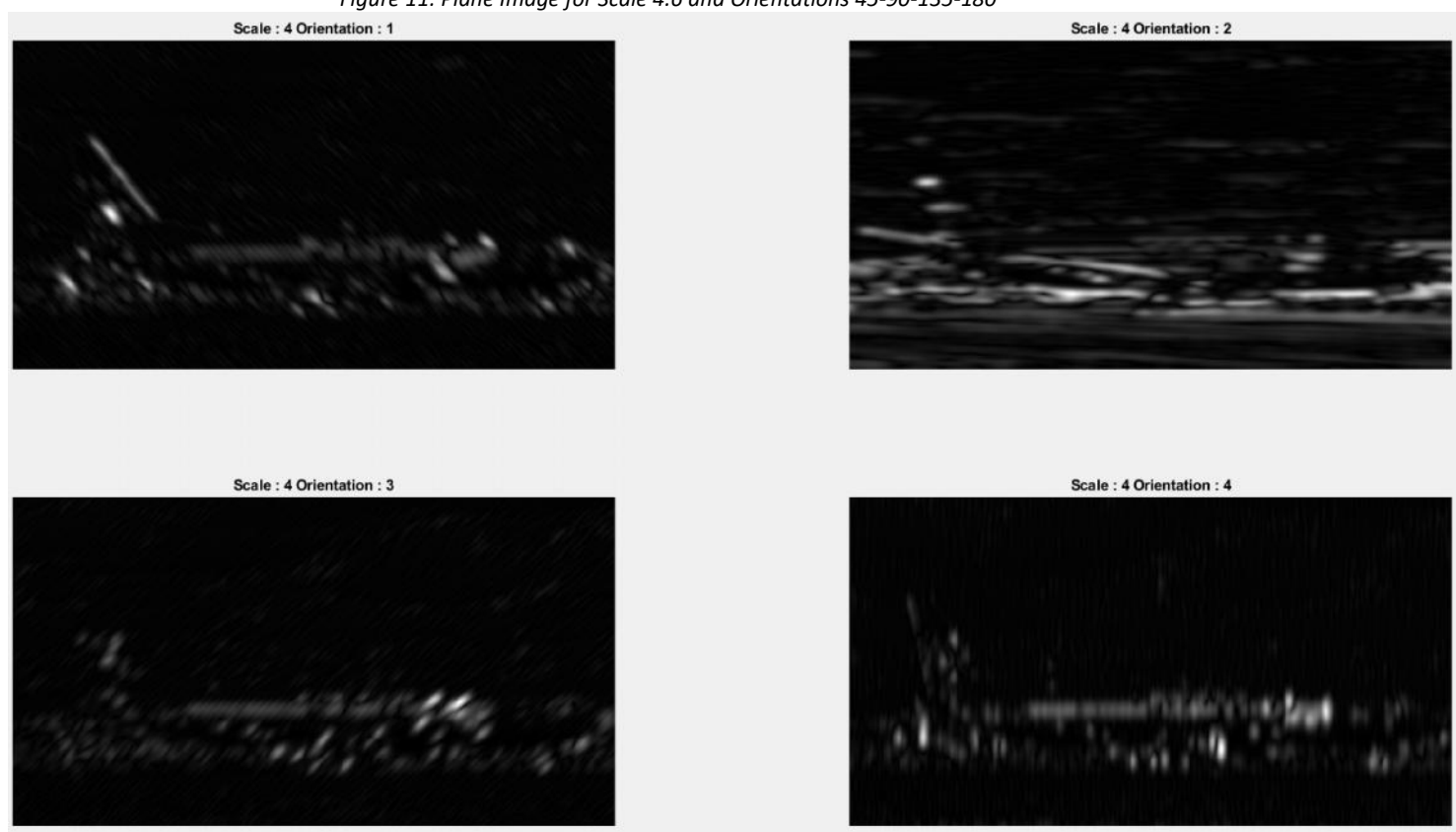


Figure 12: Plane Image for Scale 6 and Orientations 45-90-135-180

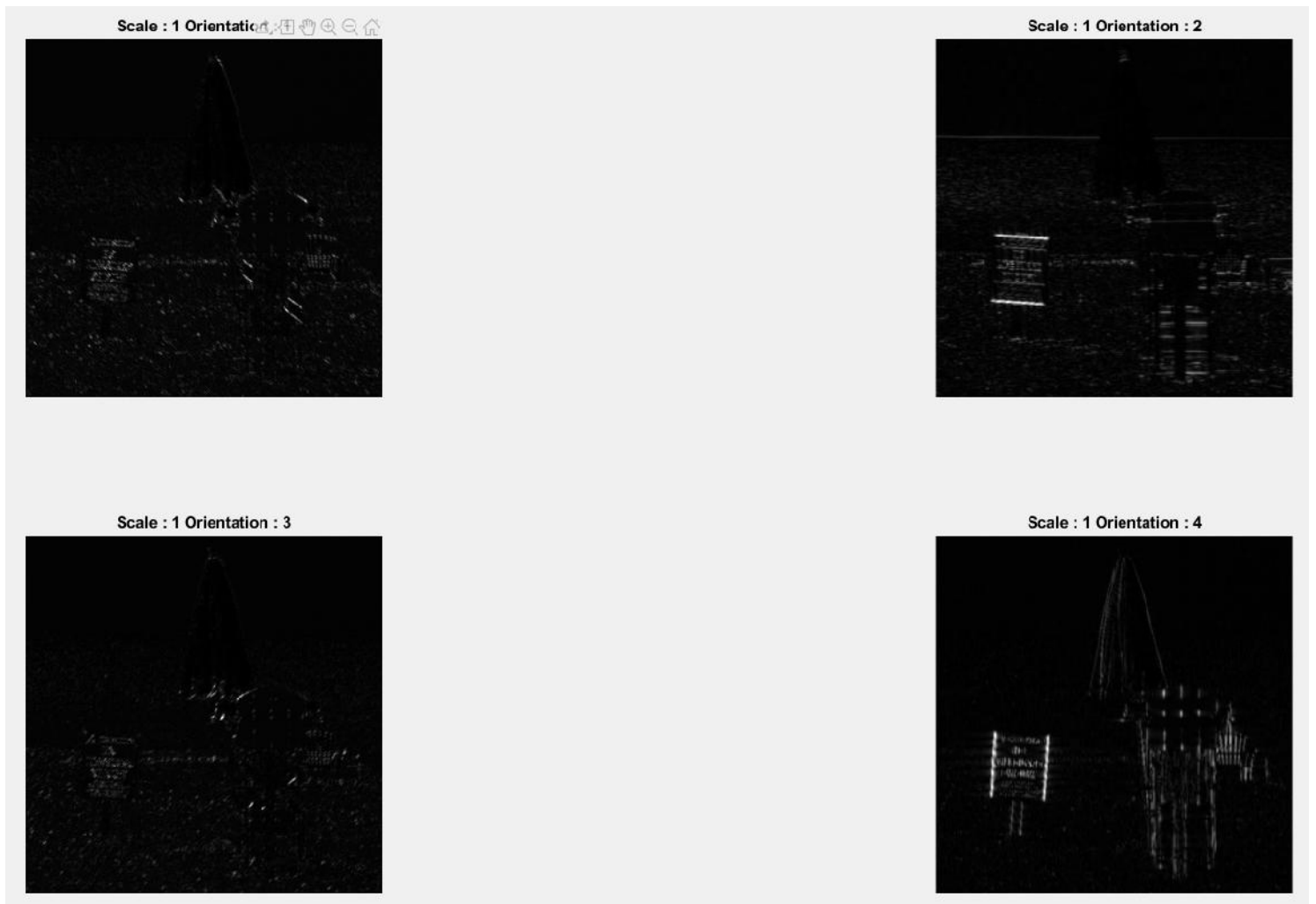
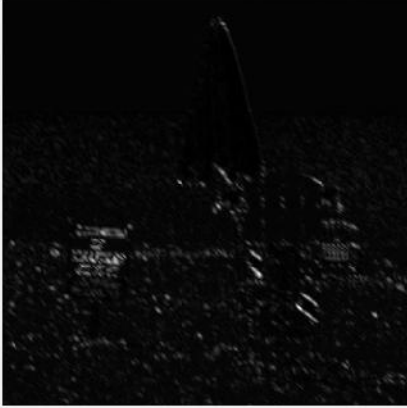
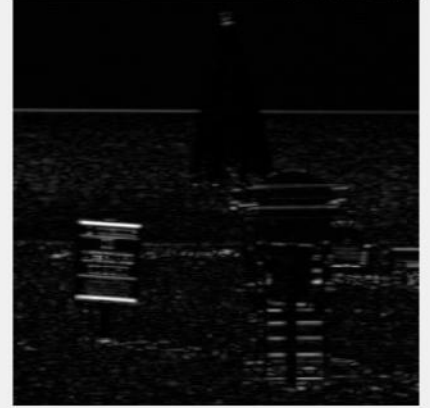


Figure 13: Beach Image for Scale 2 and Orientations 45-90-135-180

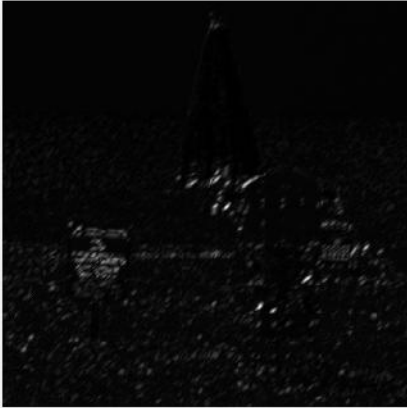
Scale : 2 Orientation : 1



Scale : 2 Orientation : 2



Scale : 2 Orientation : 3



Scale : 2 Orientation : 4

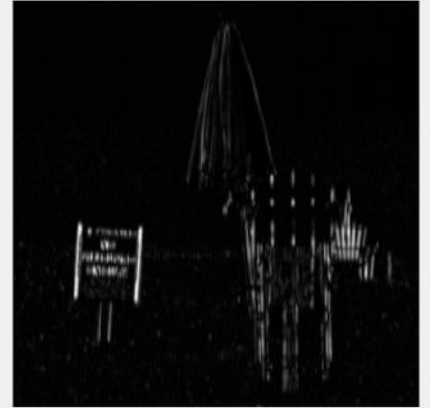
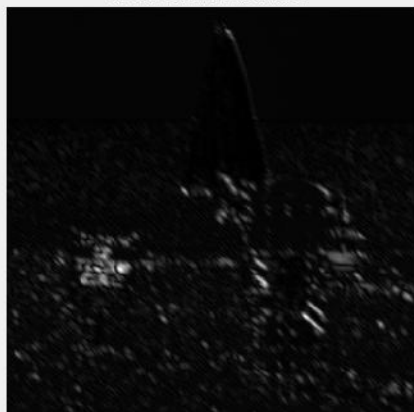
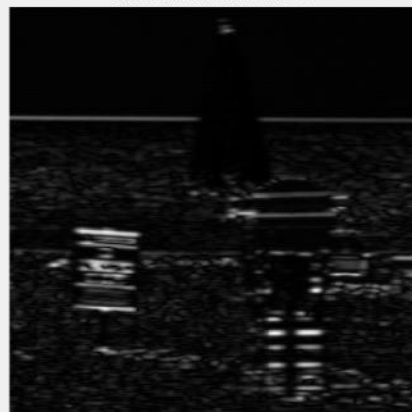


Figure 14: Beach Image for Scale 3.3 and Orientations 45-90-135-180

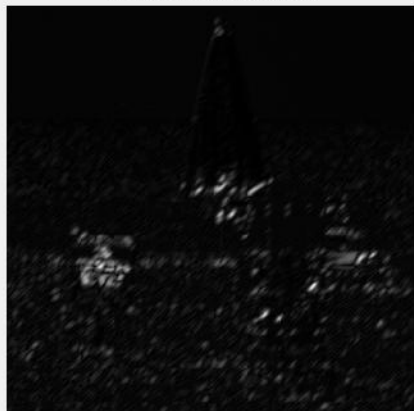
Scale : 3 Orientation : 1



Scale : 3 Orientation : 2



Scale : 3 Orientation : 3



Scale : 3 Orientation : 4

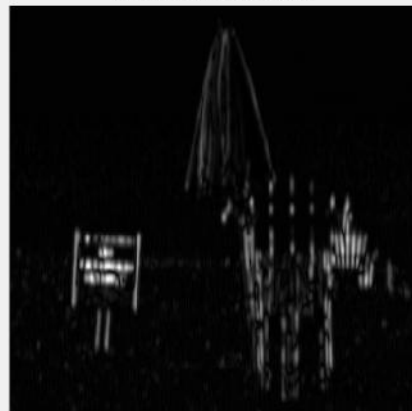
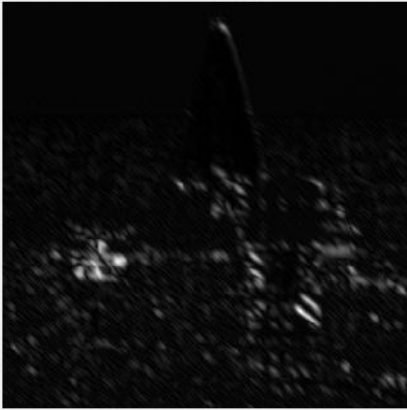
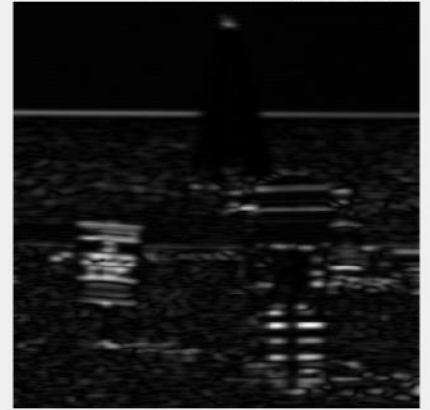


Figure 15: Beach Image for Scale 4.6 and Orientations 45-90-135-180

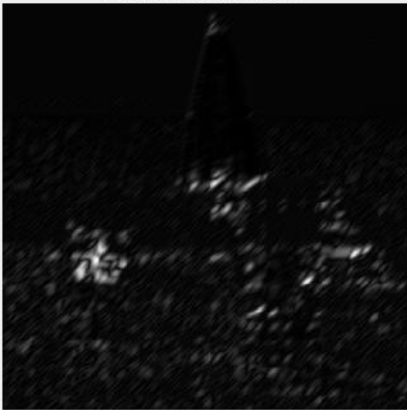
Scale : 4 Orientation : 1



Scale : 4 Orientation : 2



Scale : 4 Orientation : 3



Scale : 4 Orientation : 4

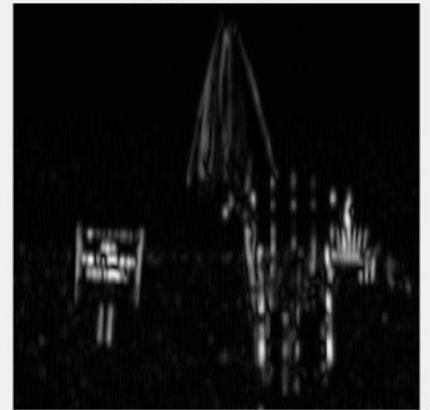


Figure 16: Beach Image for Scale 6 and Orientations 45-90-135-180

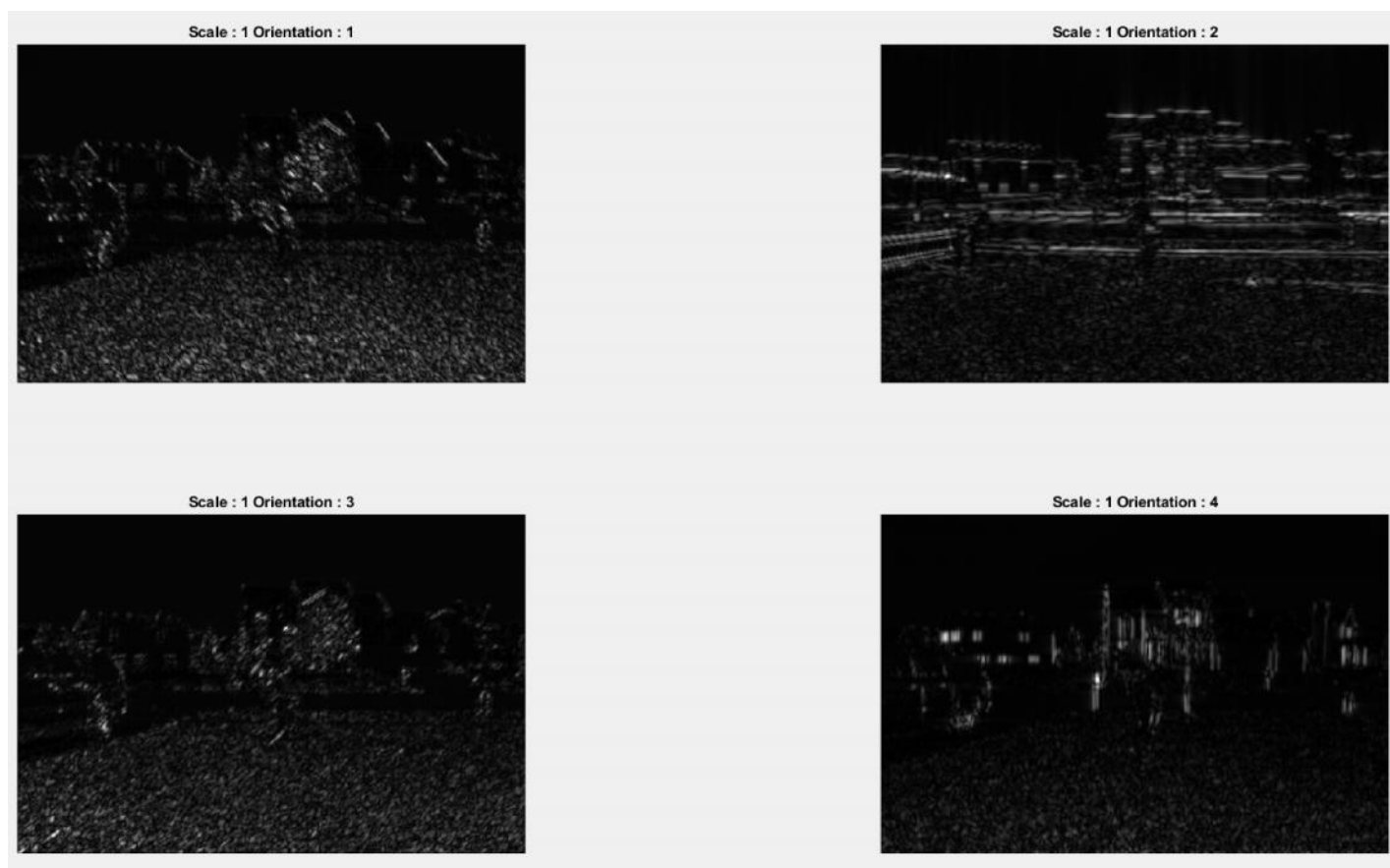


Figure 17: Houses Image for Scale 2 and Orientations 45-90-135-180

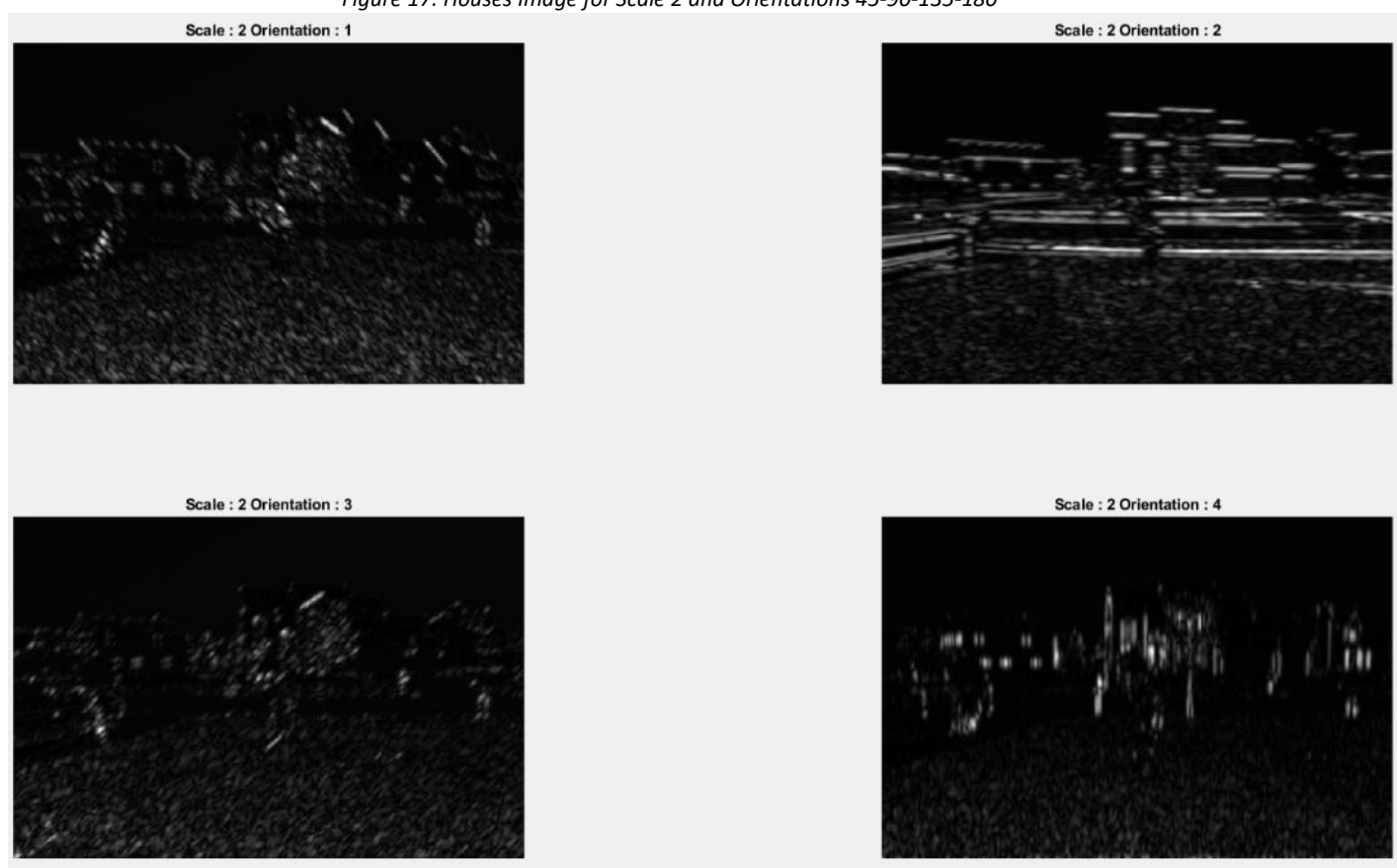


Figure 18: Houses Image for Scale 3.3 and Orientations 45-90-135-180

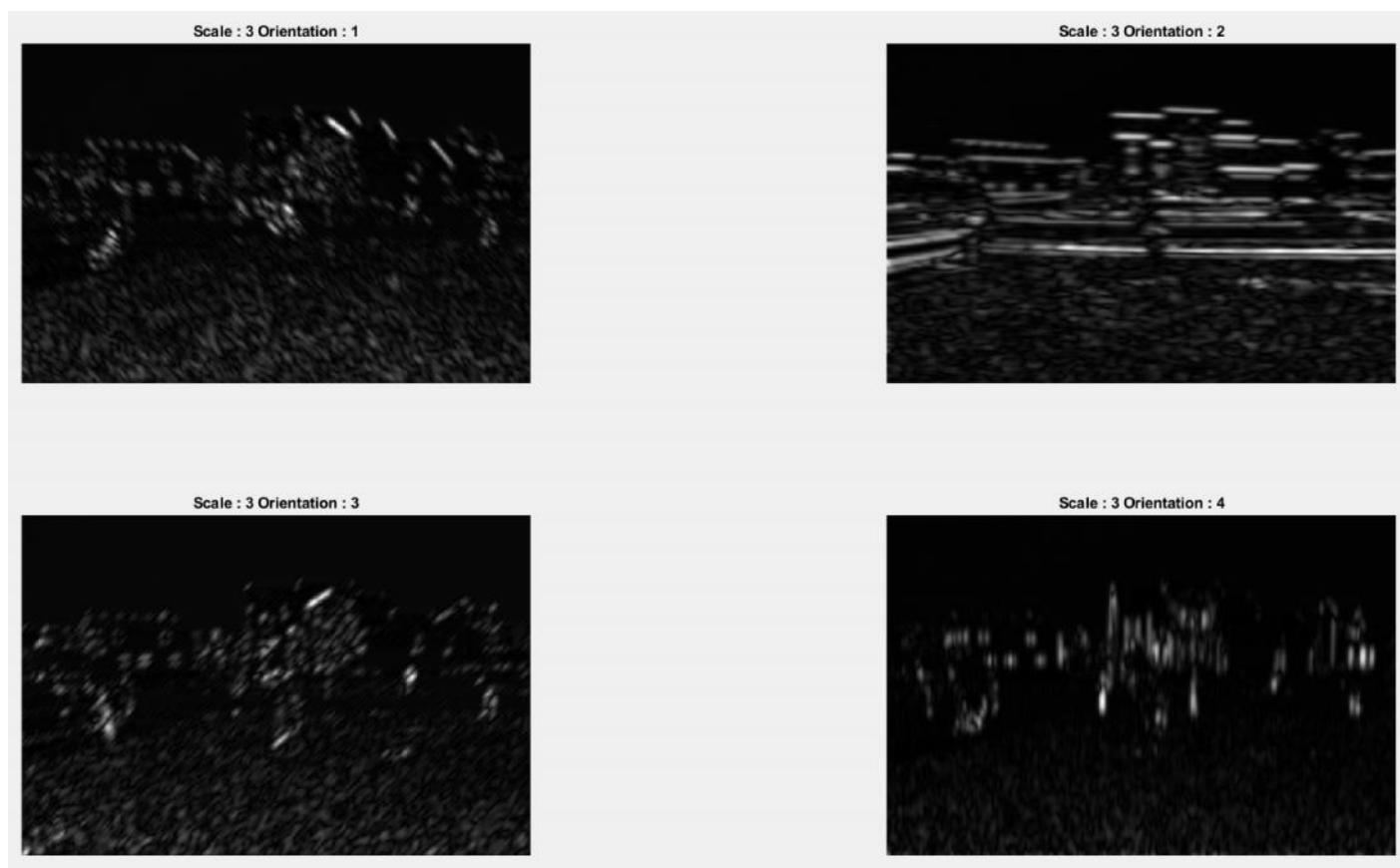


Figure 19: Houses Image for Scale 4.6 and Orientations 45-90-135-180

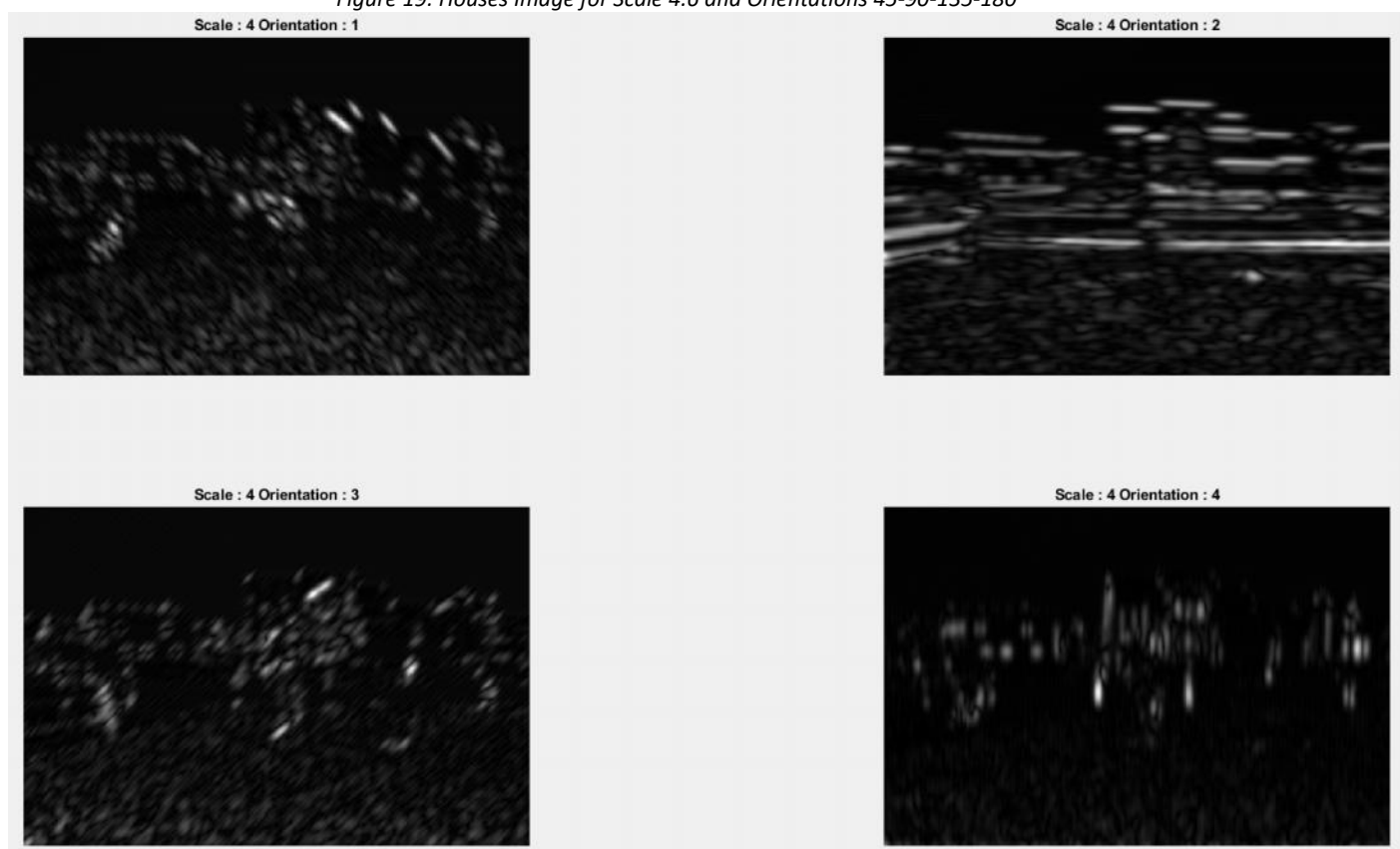


Figure 20: Houses Image for Scale 6 and Orientations 45-90-135-180

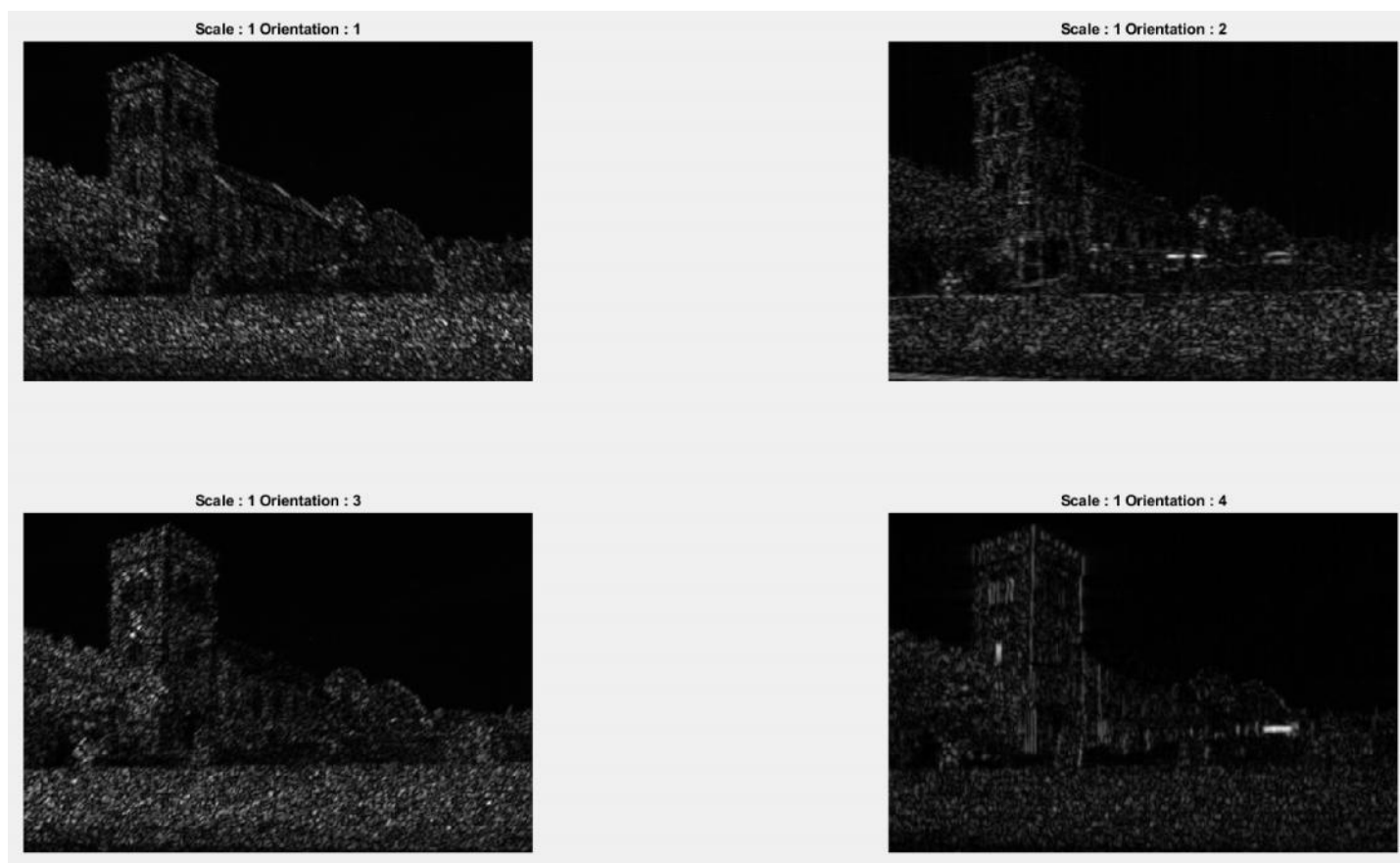


Figure 21: Cemetery Image for Scale 2 and Orientations 45-90-135-180

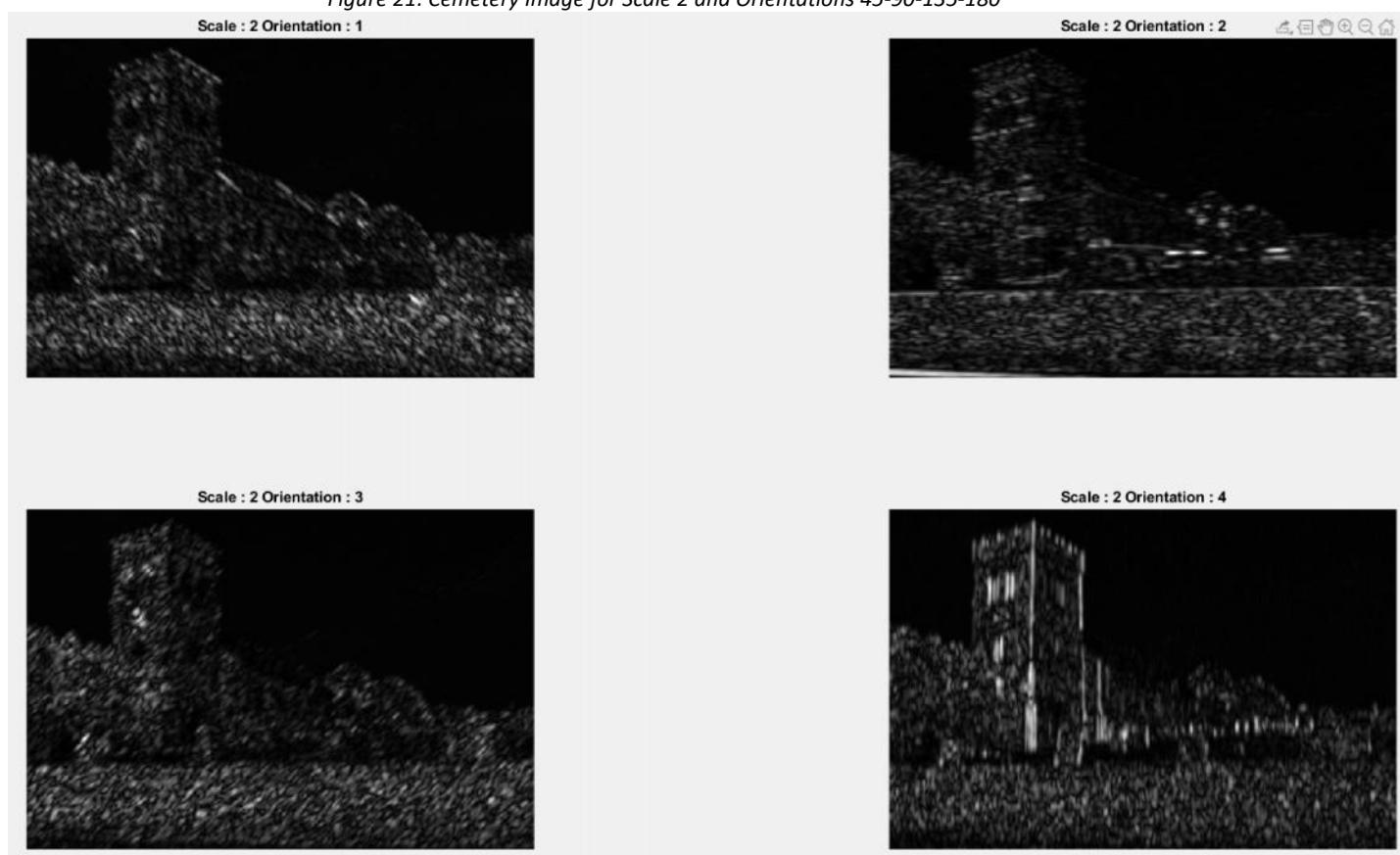


Figure 22: Cemetery Image for Scale 3.3 and Orientations 45-90-135-180

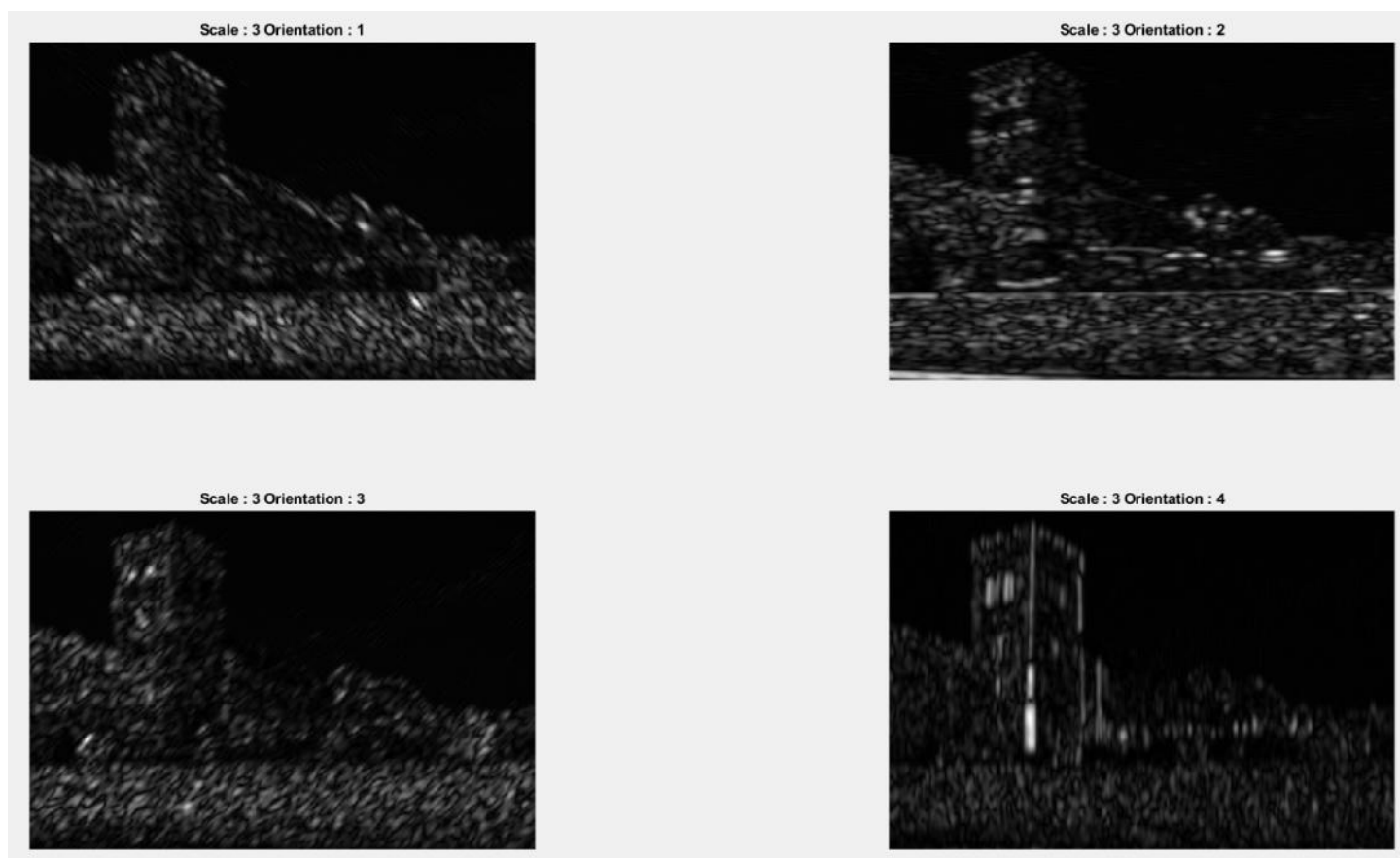


Figure 23: Cemetery Image for Scale 4.6 and Orientations 45-90-135-180

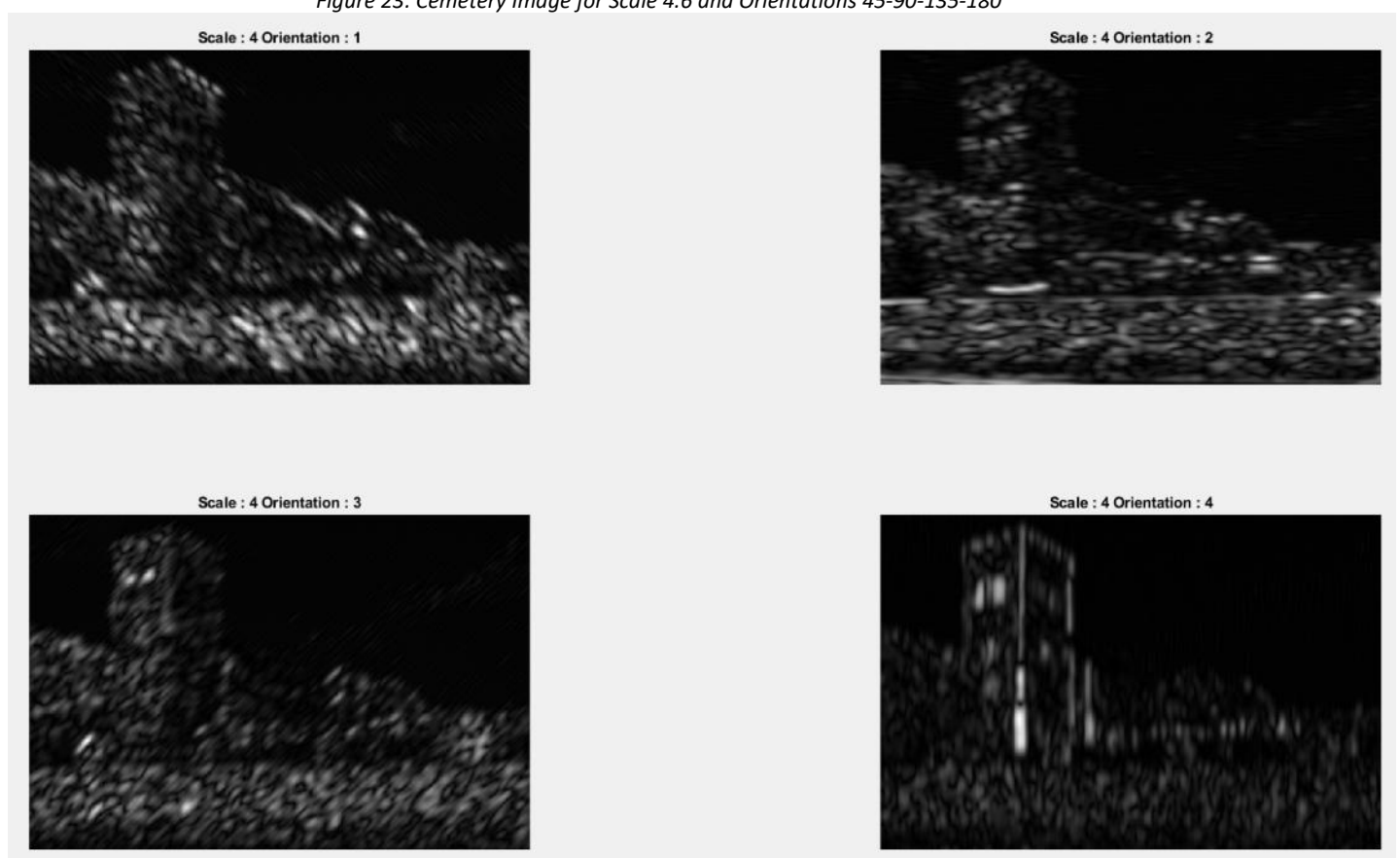


Figure 24: Cemetery Image for Scale 6 and Orientations 45-90-135-180

Scale : 1 Orientation : 1



Scale : 1 Orientation : 2



Scale : 1 Orientation : 3

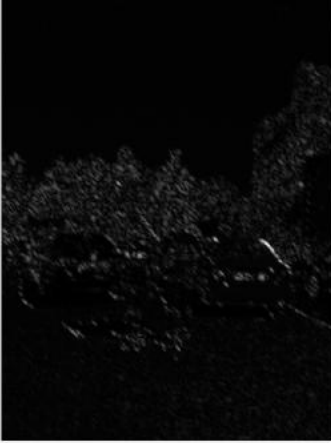


Scale : 1 Orientation : 4



Figure 25: Image I used in my First Homework for Scale 2 and Orientations 45-90-135-180

Scale : 2 Orientation : 1



Scale : 2 Orientation : 2



Scale : 2 Orientation : 3

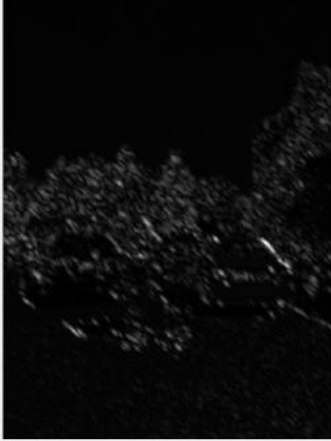


Scale : 2 Orientation : 4

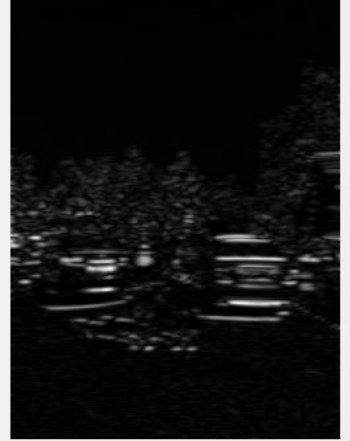


Figure 26: Image I used in my First Homework for Scale 3.3 and Orientations 45-90-135-180

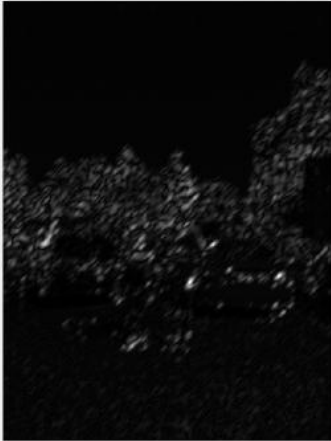
Scale : 3 Orientation : 1



Scale : 3 Orientation : 2



Scale : 3 Orientation : 3



Scale : 3 Orientation : 4

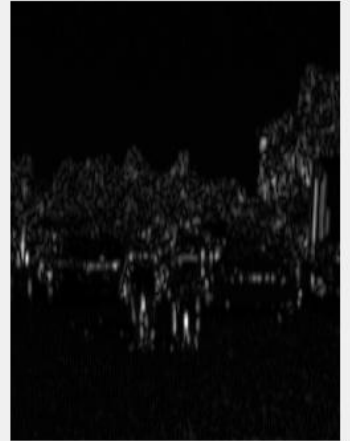


Figure 27: Image I used in my First Homework for Scale 4.6 and Orientations 45-90-135-180

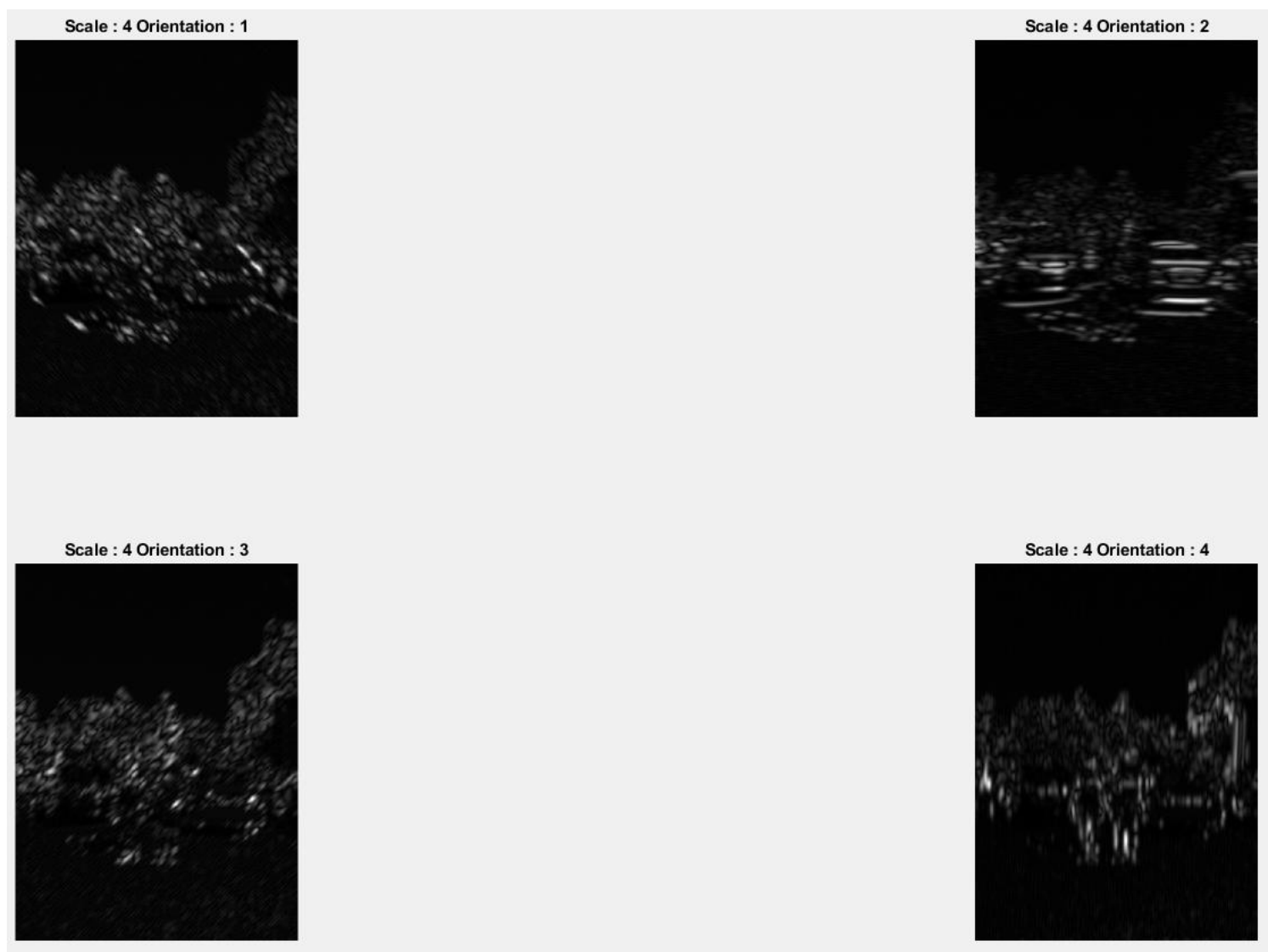


Figure 28: Image I used in my First Homework for Scale 6 and Orientations 45-90-135-180

After getting the Gabor magnitudes, I calculated their means for all superpixels for by four. The only parameter for this part was Gabor magnitudes and in the end, I had one cell with the number of labels rows and 4 for columns. Inside the columns, there were four values for each scale. Then I took the transpose of the array as I did in the colour part, and I had an array with 16 rows and the number of labels columns. I normalized the values for each scale and orientation and calculated the distance. While calculating the distances, I subtract each 16 value from each other. I put my threshold as 0.25, but as in the colour part, I did not get an optimal threshold for all images. I tested my threshold on several images, some of them were good and some of them was bad. While I optimized my threshold for one image, the other became bad, so I tried to find an average threshold for all images. After deciding on the threshold, I found my last Gabor labels.

Finally, I appended my colour and texture features to find the colour-texture labels. Since I normalized them before, I just appended them together and changed my Gabor array and method with three more array space. I changed the threshold as the other parts, but it was not much different from Gabor threshold since there are only three variables between them. The threshold of Gabor and colour was much more different. After completing all of my label arrays. I created masks by using `labeloverlay` created and applied them to my initial image. Below, you can find all of my masks applied all of the images.



Figure 29

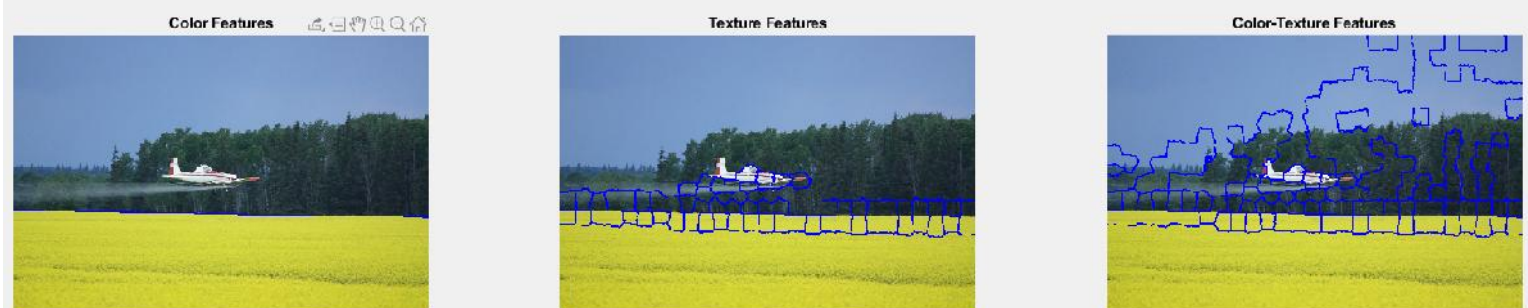


Figure 30

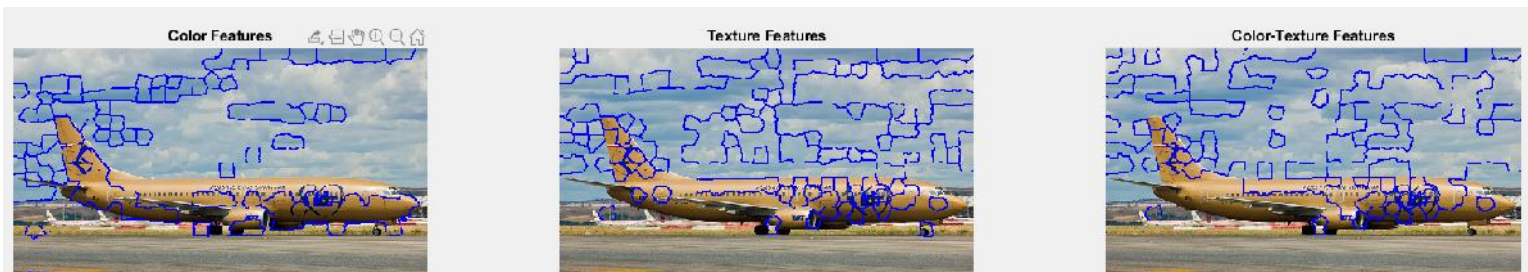
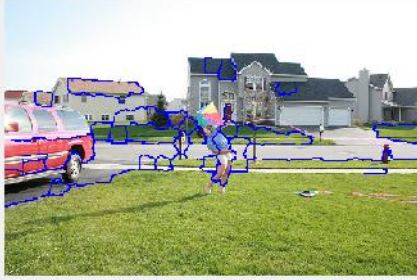


Figure 31

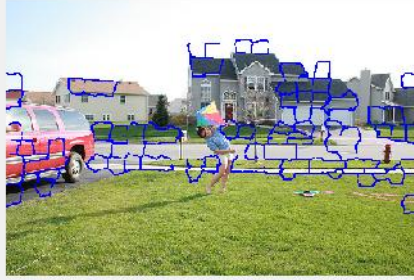


Figure 32

Color Features



Texture Features



Color-Texture Features

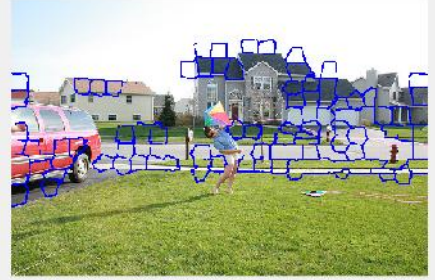
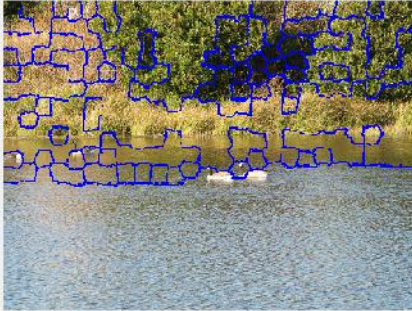


Figure 33

Color Features



Texture Features



Color-Texture Features

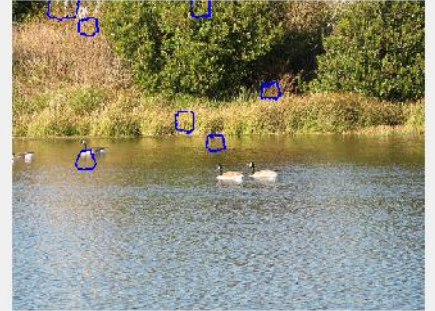


Figure 34

Color Features



Texture Features



Color-Texture Features

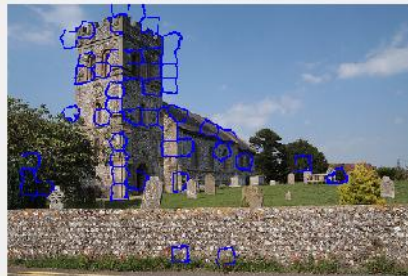


Figure 35

Color Features



Texture Features

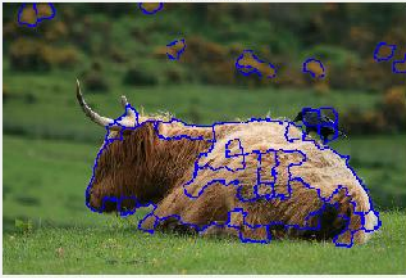


Color-Texture Features

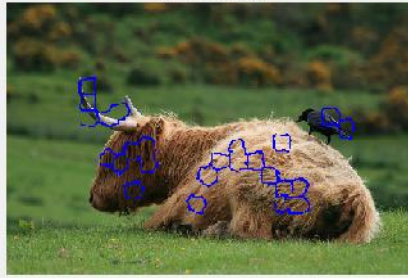


Figure 36

Color Features



Texture Features



Color-Texture Features

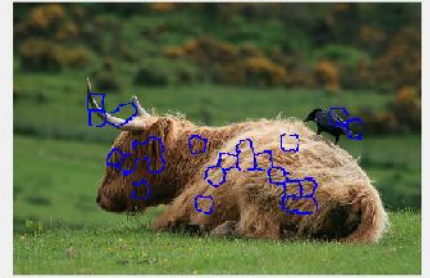
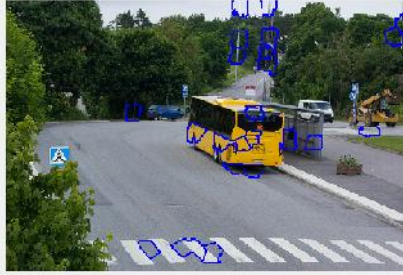


Figure 37

Color Features



Texture Features



Color-Texture Features

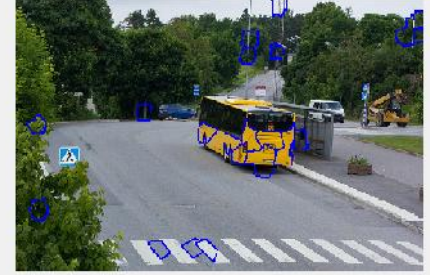
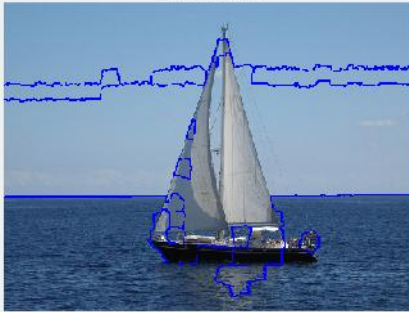
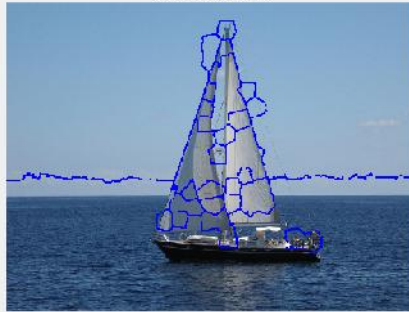


Figure 38

Color Features



Texture Features



Color-Texture Features

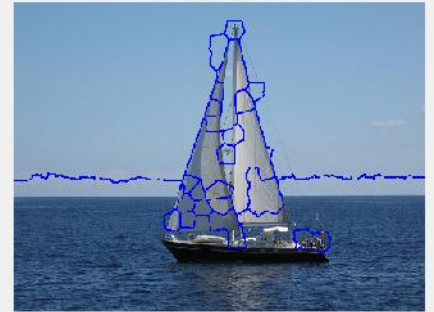
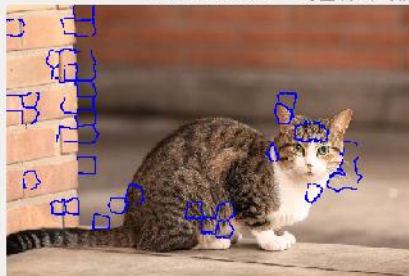


Figure 39

Color Features



Texture Features



Color-Texture Features

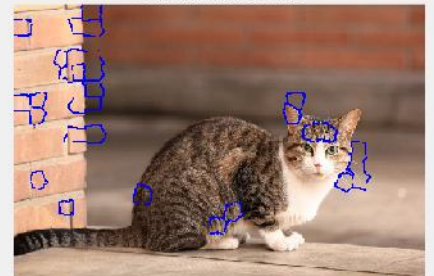


Figure 40

Color Features



Texture Features



Color-Texture Features



Figure 41

Color Features



Texture Features



Color-Texture Features

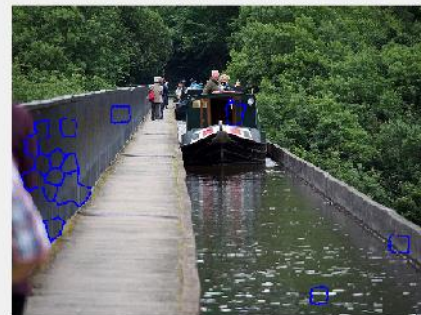
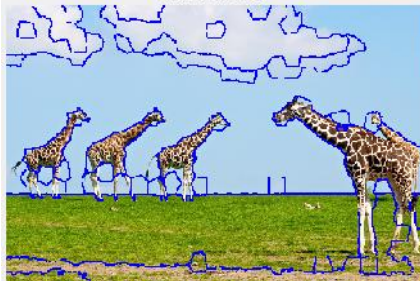
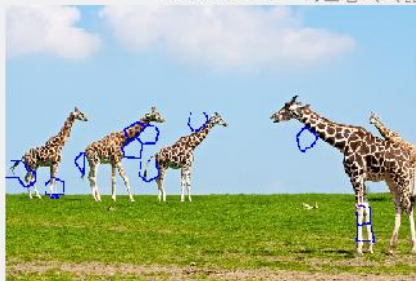


Figure 42

Color Features



Texture Features



Color-Texture Features

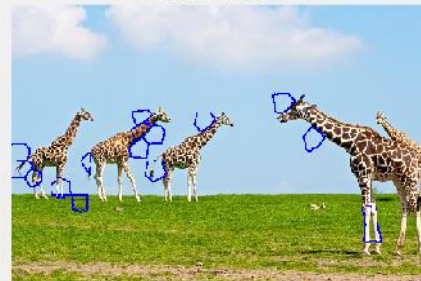
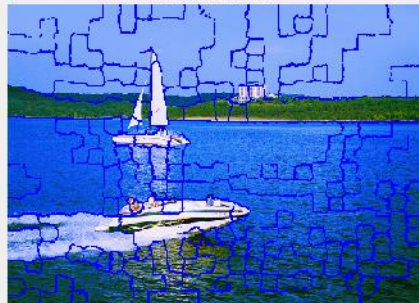


Figure 43

Color Features



Texture Features



Color-Texture Features

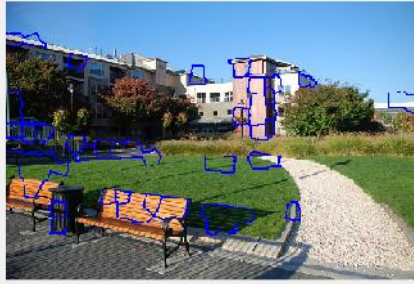


Figure 44

Color Features



Texture Features



Color-Texture Features

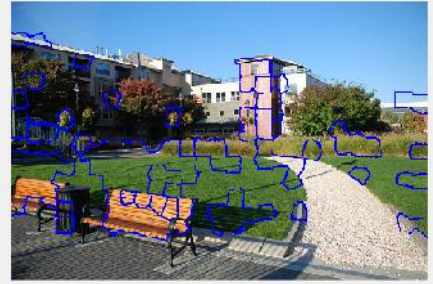
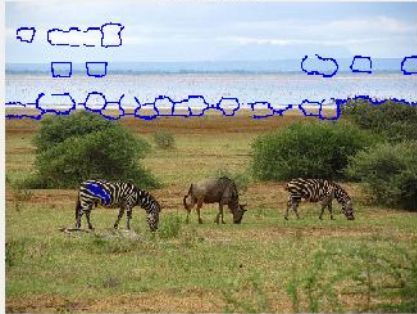
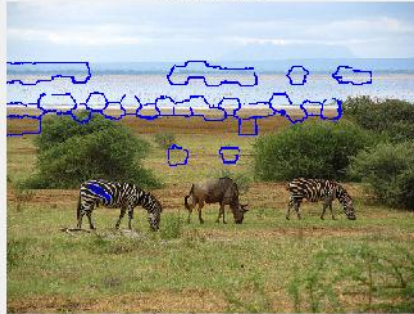


Figure 45

Color Features



Texture Features



Color-Texture Features

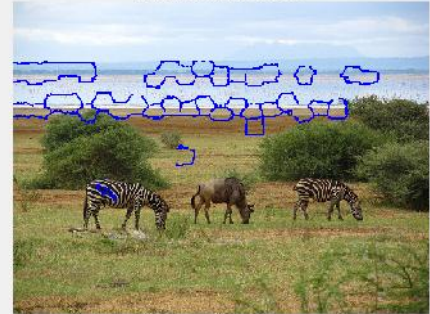
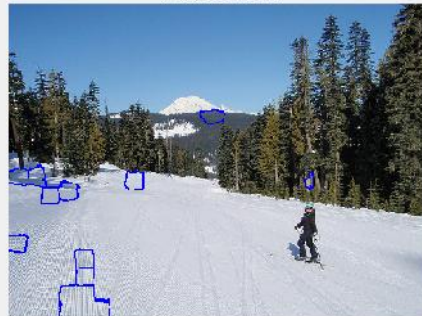


Figure 46

Color Features



Texture Features



Color-Texture Features

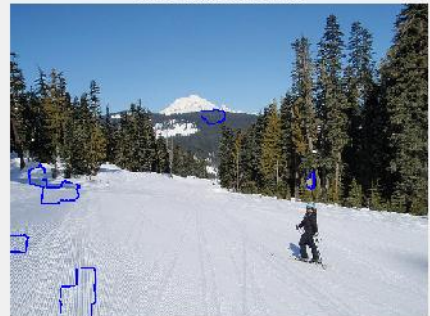
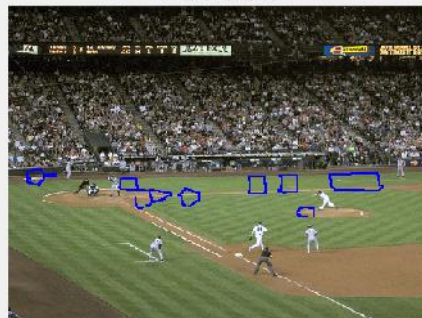


Figure 47

Color Features



Texture Features



Color-Texture Features



Figure 48



Figure 49

Discussion

As I mentioned while describing the functions, I could not find a good threshold for all of the images. For the images that have a clearly distinguishable colour difference, the colour filtering was working fine. In the images such as there are sky and grass below, it put a good line between them such as Figure 45. However, in the images like red bus (Figure 29), since the bus is red and there are also some different colours like windows, it could not put a good line between bus and windows, or in Figure 44, since there are lots of waves in the sea, and we are taking the mean of the colours, probably the mean changed because of the colour of the waves and colour mask was not good enough. It could have been better if I select my number of pixels higher, but then I thought I could also lose some borders if I would do that. Therefore, I did not change the number of superpixels. For Gabor filtering, it also put a good border in lines that it can detect, however, since it cannot differentiate colours, in some images that colour filtering was good, Gabor filtering was not good. I think colour filtering was overall better in my implementation. Moreover, since there are 16 variables in the Gabor mask, I could not understand if I did the implementation correct. There were sixteen variables in the Gabor filtering, so I think it lost many details while I calculated the distance. I think if I would just apply Gabor filtering in four scales and one orientation such as 0, The borders for that angle would be much better. However, since I combined all the angles, and calculated their distances together, I lost many details. For the colour-texture filter, since it is the combined version of both colour and Gabor, it was better than Gabor. However, I think since three average colour value distanced with sixteen Gabor value, colour means also lost many details and it was worse than the colour filter. You can see from the example results that there was not much difference between the texture and colour-texture mask since texture values have a higher number. According to my observation from my results, the colour mask was

nearly always better than the others. This could be because of my implementation, but I could not find any better result. Also, my results generally were not that good, so there could be something wrong with my implementation.