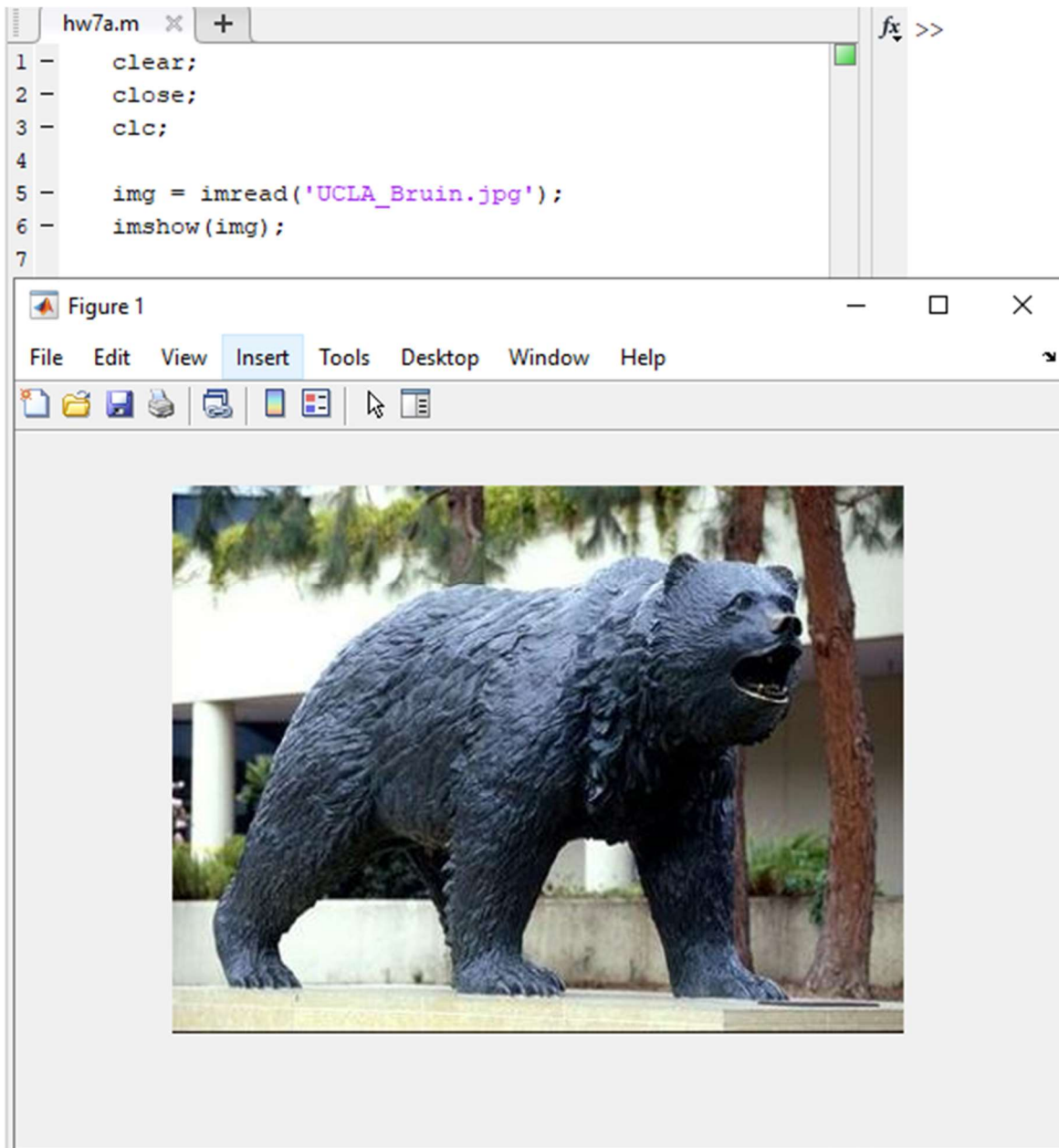
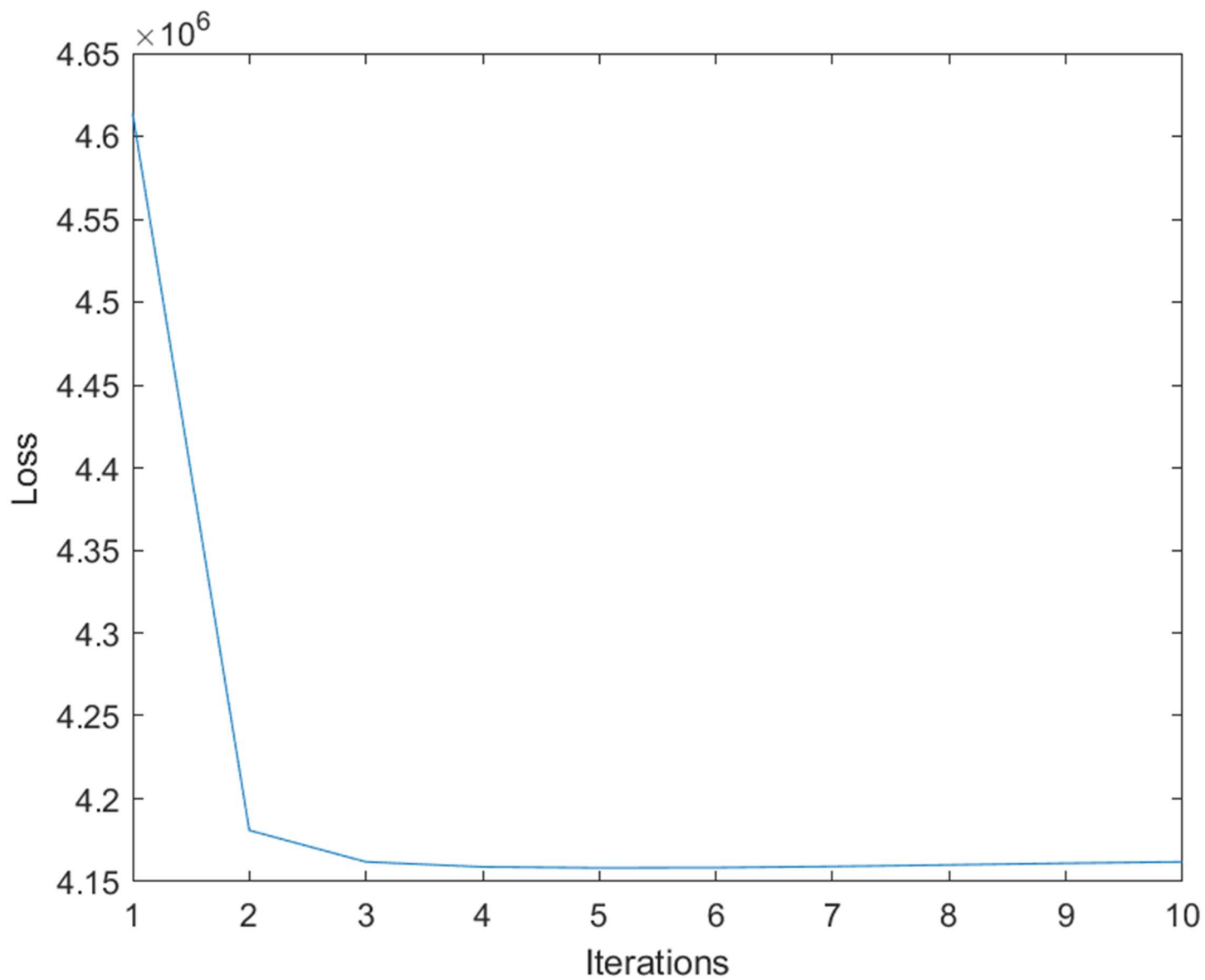


5.

a.



b.



The algorithm converges after about 4-5 iterations and after that, the loss function doesn't change much.

Code:

```
1 clear;
2 close;
3 %clc;
4
5 img = double(imread('UCLA_Bruin.jpg'));
6 K = 4;
7 numIter = 10;
8 u(1,:) = double(squeeze(img(1,1,:)))';
9 numMeans = 1;
10
11 for k = 2:K
12     for i = 1:numMeans
13         %dist of all pixels to ith mu (u)
14         dist(:,:,i) = sqrt((img(:,:,1) - u(i,1)).^2 + (img(:,:,2) - u(i,2)).^2 + (img(:,:,3) - u(i,3)).^2);
15     end
16     % now know dist of every pixel to every mu (that currently exists)
17
18     % the min dist for each pixel (to the closest u)
19     minDists = min(dist,[],3);
20
21     [M,I] = max(minDists);
22     [N,J] = max(M); %N is largest elem of minDists
23     Z = I(J);
24     if(N ~= minDists(Z,J))
25         disp("ERROR!");
26     end
27     %dist(Z,J) is the max dist
28     %img(Z,J,:) is the RGB for that pixel
29
30     newMeanIndex = numMeans+1;
31     u(newMeanIndex,1) = img(Z,J,1);
32     u(newMeanIndex,2) = img(Z,J,2);
33     u(newMeanIndex,3) = img(Z,J,3);
34     numMeans = numMeans + 1;
35 end
36 % now initialized (k mu's set by furthest-first rule)
37
38 %each iter do assignment and re-estimation of center of cluster
39 %and also calc loss func for tracking the alg
40 Loss = zeros(1,numIter);
41 for iter = 1:numIter
42     %assignment
43     for k = 1:K
44         dist(:,:,k) = sqrt((img(:,:,1) - u(k,1)).^2 + (img(:,:,2) - u(k,2)).^2 + (img(:,:,3) - u(k,3)).^2);
45     end
46     for i = 1:size(img,1)
47         for j = 1:size(img,2)
48             [M,I] = min(dist(i,j,:));
49             assign(i,j) = I;
50         end
51     end
52
53     %re-estimation
54     % pts(i,:) is the sum of the RGB elem of all pts in img w/ assign() = i
55     pts = zeros(K,3);
56     numPts = zeros(1,4);
57     for i = 1:size(img,1)
58         for j = 1:size(img,2)
59             tmp = assign(i,j);
60             pts(tmp,:) = pts(tmp,:) + squeeze(img(i,j,:))';
61             numPts(tmp) = numPts(tmp) + 1;
62         end
63     end
64
65     for k = 1:K
66         u(k,:) = pts(k,:)/numPts(k);
67     end
68
69     %loss func
70     for i = 1:size(img,1)
71         for j = 1:size(img,2)
72             tmp = assign(i,j);
73             Loss(iter) = Loss(iter) + norm(squeeze(img(i,j,:))' - u(tmp,:));
74         end
75     end
76 end
77
78 plot([1:numIter], Loss);
79 xlabel("Iterations")
80 ylabel("Loss")
```

c.

K = 4:



Final value of objective function:  $4.1619 \times 10^6$

K = 8:



Final value of objective function:  $2.827 \times 10^6$



K = 16:



Final value of objective function:  $1.8979 \times 10^6$

- Increasing K increases the quality of the end image. This is because there are more colors that can be used. So for K = 4, each pixel gets assigned to one of 4 colors which will have less detail/quality than if they were assigned to one of 16 colors.

d.

300x400 = 120,000 pixels, each taking 3\*8 bits

120,000\*24 bits = 2,880,000 bits = 2812.5 Kilo bits (2,880,000/1024)

→ **2,880,000 bits to store original image**

Data Compression Ratio = uncompressed size / compressed size

New:

K = 4:

- 2 bits for each pixel to store the index of which cluster =  $\log(\text{base } 2)4$
- + 8 bits for each cluster \* 4 clusters = 32 bits
- 120,000\*2 = 240,000 + 32 = 240,032 bits for K = 4
- Compression Ratio = 2,880,000/240,032 = 11.998

K = 8:

- 3 bits for each pixel to store the index of which cluster =  $\log(\text{base } 2)8$
- + 8 bits for each cluster \* 8 clusters = 64 bits
- 120,000\*3 = 360,000 + 64 = 360,064 bits for K = 8
- Compression Ratio = 2,880,000/360,064 = 7.999

K = 16:

- 4 bits for each pixel to store the index of which cluster =  $\log(\text{base } 2)16$
- + 8 bits for each cluster \* 16 clusters = 128 bits
- 120,000\*4 = 480,000 + 128 = 480,128 bits for K = 16
- Compression Ratio = 2,880,000/480,128 = 5.998

K	Bits	Compression Ratio
4	240,032	11.998
8	360,064	7.999
16	480,128	5.998