

Math 123 Homework 6

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Question 1

(a)

$$vol(C) = \sum_{\substack{x_i \in C \\ x_j \in V}} W_{ij} = \sum_{\substack{x_i \in C \\ x_j \in N(x_i)}} W_{ij} \quad (1)$$

$N(x_i)$ is the set of neighborhood of x_i , $(x_j | w_{ij} \neq 0, \text{ for } x_i)$

$$D_{ii} = \sum_{x_j \in N(x_i)} W_{ij} \quad (2)$$

$$(Df^C)_i = \begin{cases} -\sum_{x_j \in N(x_i)} W_{ij} \sqrt{vol(\bar{C})/vol(C)} & x_i \in C \\ \sum_{x_j \in N(x_i)} W_{ij} \sqrt{vol(C)/vol(\bar{C})} & x_i \in \bar{C} \end{cases} \quad (3)$$

$$\begin{aligned} \langle Df^C \cdot \mathbb{1} \rangle &= \sum_{x_i \in V} (Df^C)_i = \sum_{x_i \in C} (Df^C)_i + \sum_{x_i \in \bar{C}} (Df^C)_i \\ &\stackrel{(3)}{=} -\sum_{\substack{x_i \in C \\ x_j \in V}} W_{ij} \sqrt{vol(\bar{C})/vol(C)} + \sum_{\substack{x_i \in \bar{C} \\ x_j \in V}} W_{ij} \sqrt{vol(C)/vol(\bar{C})} \\ &= -\sqrt{vol(\bar{C})/vol(C)} \cdot \sum_{\substack{x_i \in C \\ x_j \in V}} W_{ij} + \sqrt{vol(C)/vol(\bar{C})} \cdot \sum_{\substack{x_i \in \bar{C} \\ x_j \in V}} W_{ij} \quad (4) \\ &\stackrel{(1)}{=} -\sqrt{vol(\bar{C})/vol(C)} \cdot vol(C) + \sqrt{vol(C)/vol(\bar{C})} \cdot vol(\bar{C}) \\ &= -\sqrt{vol(\bar{C})vol(C)} + \sqrt{vol(C)vol(\bar{C})} \\ &= 0 \end{aligned}$$

(b)

$$\begin{aligned}
(f^C)^T D f^C &= \sum_{x_i \in V} f^C(x_i) \left(\sum_{x_j \in N(x_i)} W_{ij} \right) f^C(x_i) \\
&= \sum_{x_i \in V} (f^C(x_i))^2 \left(\sum_{x_j \in N(x_i)} W_{ij} \right) \\
&= \sum_{x_i \in C} (f^C(x_i))^2 \left(\sum_{x_j \in N(x_i)} W_{ij} \right) + \sum_{x_i \in \bar{C}} (f^C(x_i))^2 \left(\sum_{x_j \in N(x_i)} W_{ij} \right) \\
&= \text{vol}(C) \frac{\text{vol}(\bar{C})}{\text{vol}(C)} + \text{vol}(\bar{C}) \frac{\text{vol}(C)}{\text{vol}(\bar{C})} \\
&= \text{vol}(\bar{C}) + \text{vol}(C) \\
&= \sum_{\substack{x_i \in \bar{C} \\ x_j \in V}} W_{ij} + \sum_{\substack{x_i \in C \\ x_j \in V}} W_{ij} = \sum_{\substack{x_i \in V \\ x_j \in V}} W_{ij} \\
&= \text{vol}(V)
\end{aligned} \tag{5}$$

(c)

$$(f^C)^T L f^C = (f^C)^T (D - W) f^C = (f^C)^T D f^C - (f^C)^T W f^C \tag{6}$$

We know that

$$(f^C)^T D f^C = \text{vol}(V) \tag{7}$$

And

$$\begin{aligned}
(f^C)^T W f^C &= \sum_{\substack{x_i \in V \\ x_j \in V}} W_{ij} f_i^C f_j^C \\
&= \sum_{\substack{x_i \in C \\ x_j \in C}} W_{ij} f_i^C f_j^C + \sum_{\substack{x_i \in \bar{C} \\ x_j \in \bar{C}}} W_{ij} f_i^C f_j^C + \sum_{\substack{x_i \in C \\ x_j \in \bar{C}}} W_{ij} f_i^C f_j^C + \sum_{\substack{x_i \in \bar{C} \\ x_j \in C}} W_{ij} f_i^C f_j^C
\end{aligned} \tag{8}$$

Notice that

$$f_i^C f_j^C = \begin{cases} \frac{\text{vol}(\bar{C})}{\text{vol}(C)} & x_i \in C \text{ and } x_j \in C \\ \frac{\text{vol}(C)}{\text{vol}(\bar{C})} & x_i \in \bar{C} \text{ and } x_j \in \bar{C} \\ -1 & x_i \in \bar{C} \text{ and } x_j \in C, \text{ or } x_i \in C \text{ and } x_j \in \bar{C} \end{cases}$$

, and we define $\sum_{\substack{x_i \in C \\ x_j \in \bar{C}}} W_{ij} = \sum_{\substack{x_i \in \bar{C} \\ x_j \in C}} W_{ij} =: \frac{1}{2} \text{vol}(B)$.

Then (8) turns out to be

$$\begin{aligned}
(8) &= \sum_{\substack{x_i \in C \\ x_j \in C}} W_{ij} \frac{vol(\bar{C})}{vol(C)} + \sum_{\substack{x_i \in \bar{C} \\ x_j \in \bar{C}}} W_{ij} \frac{vol(C)}{vol(\bar{C})} - \sum_{\substack{x_i \in C \\ x_j \in \bar{C}}} W_{ij} - \sum_{\substack{x_i \in \bar{C} \\ x_j \in C}} W_{ij} \\
&= \frac{vol(\bar{C})}{vol(C)} \sum_{\substack{x_i \in C \\ x_j \in C}} W_{ij} + \frac{vol(C)}{vol(\bar{C})} \sum_{\substack{x_i \in \bar{C} \\ x_j \in \bar{C}}} W_{ij} - \frac{1}{2} vol(B) - \frac{1}{2} vol(B) \\
&= \frac{vol(\bar{C})}{vol(C)} \left(\sum_{\substack{x_i \in C \\ x_j \in V}} W_{ij} - \sum_{\substack{x_i \in C \\ x_j \in C}} W_{ij} \right) + \frac{vol(C)}{vol(\bar{C})} \left(\sum_{\substack{x_i \in \bar{C} \\ x_j \in V}} W_{ij} - \sum_{\substack{x_i \in \bar{C} \\ x_j \in C}} W_{ij} \right) - vol(B) \\
&= \frac{vol(\bar{C})}{vol(C)} \left(vol(C) - \frac{1}{2} vol(B) \right) + \frac{vol(C)}{vol(\bar{C})} \left(vol(\bar{C}) - \frac{1}{2} vol(B) \right) - vol(B) \\
&= vol(\bar{C}) - \frac{1}{2} vol(B) \frac{vol(\bar{C})}{vol(C)} + vol(C) - \frac{1}{2} vol(B) \frac{vol(C)}{vol(\bar{C})} - vol(B) \\
&= vol(V) - \frac{1}{2} vol(B) \left(\frac{vol(\bar{C})}{vol(C)} + \frac{vol(C)}{vol(\bar{C})} + 2 \right) \\
&= vol(V) - \frac{1}{2} vol(B) \left(\sqrt{\frac{vol(\bar{C})}{vol(C)}} + \sqrt{\frac{vol(C)}{vol(\bar{C})}} \right)^2 \\
&= vol(V) - \frac{1}{2} vol(B) \frac{(\sqrt{vol(\bar{C})} \sqrt{vol(\bar{C})} + \sqrt{vol(C)} \sqrt{vol(C)})^2}{vol(\bar{C}) vol(C)} \\
&= vol(V) - \frac{1}{2} vol(B) \frac{vol(V)^2}{vol(\bar{C}) vol(C)} \\
&= vol(V) - \frac{1}{2} vol(B) vol(V) \frac{vol(V)}{vol(\bar{C}) vol(C)} \\
&= vol(V) - \frac{1}{2} vol(B) vol(V) \frac{vol(\bar{C}) + vol(C)}{vol(\bar{C}) vol(C)} \\
&= vol(V) - \frac{1}{2} vol(B) vol(V) \left(\frac{1}{vol(\bar{C})} + \frac{1}{vol(C)} \right) \\
&= vol(V) - vol(V) Ncut(C, \bar{C})
\end{aligned} \tag{9}$$

, where $Ncut(C, \bar{C}) = \sum_{\substack{x_i \in C \\ x_j \in \bar{C}}} W_{ij} \left(\frac{1}{vol(\bar{C})} + \frac{1}{vol(C)} \right) = \frac{1}{2} vol(B) \left(\frac{1}{vol(\bar{C})} + \frac{1}{vol(C)} \right)$

Substitute (7) and (9) into (6), we have

$$\begin{aligned}
(f^C)^T L f^C &= (f^C)^T (D - W) f^C = (f^C)^T D f^C - (f^C)^T W f^C \\
&= vol(V) - (vol(V) - vol(V) Ncut(C, \bar{C})) \\
&= vol(V) Ncut(C, \bar{C})
\end{aligned} \tag{10}$$

Question 2

(a) Because

$$L_{ij} = \begin{cases} -W_{ij} & i \neq j \\ \sum_{\substack{x_j \in V \\ i \neq j}} W_{ij} & i = j \end{cases}$$

, and

$$\lim_{\sigma \rightarrow 0} W_{ij} = \lim_{\sigma \rightarrow 0} e^{-\frac{\|x_i - x_j\|_2^2}{\sigma^2}} = 0$$

, then

$$\lim_{\sigma \rightarrow 0} L_{ij} = 0$$

Therefore, all eigenvalues are 0s and any $v \in \mathbb{R}^n / \vec{0}$ are the eigenvalues.

(b) Since

$$\lim_{\sigma \rightarrow \infty} W_{ij} = \lim_{\sigma \rightarrow \infty} e^{-\frac{\|x_i - x_j\|_2^2}{\sigma^2}} = 1$$

, then

$$\lim_{\sigma \rightarrow \infty} L_{ij} = \begin{cases} -1, & i \neq j \\ n - 1, & i = j \end{cases}$$

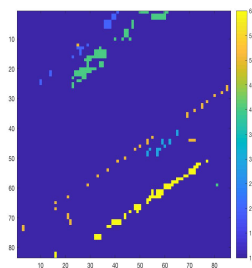
It is equivalent to a unweighted complete graph. Therefore, it has one eigenvalue at 0, and other eigenvalues at n . The eigenvector of 0 is $\mathbb{1}$ and all other eigenvectors are distinct.

Question 3

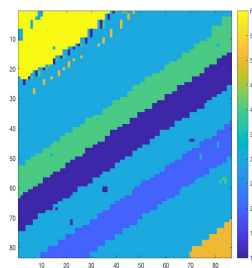
(a) We choose $\sigma = 50000$ here and $k = 6$. When the knn is about 15 to 25, the clusters look similar to groundtruth.

Firstly, it is clear that our result shows less information than groundtruth, because the groundtruth has 14 categories, while we only have 6 clusters (in eigenspace) for kmean method.

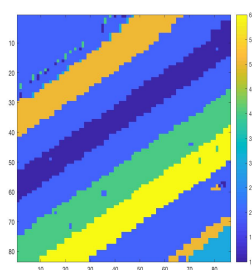
Secondly, When knn is too small, the weighted degree of different nodes varies too much. Some points in "very dense" area become outliers in eigenspace, which results in kmean producing poor clusters (lots of clusters are outliers only).



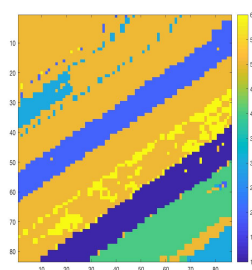
(a) knn=5



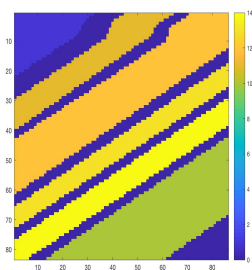
(b) knn=15



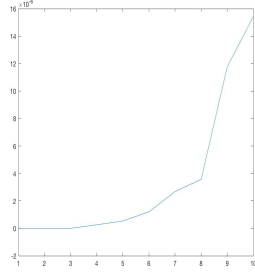
(a) knn=25



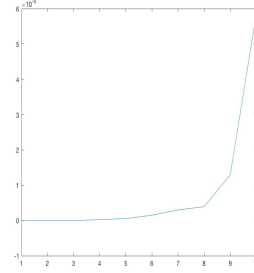
(b) knn=50



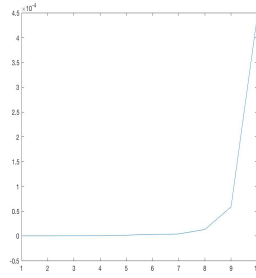
(c) groundtruth



(a) $\sigma=500$



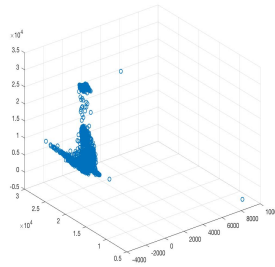
(b) $\sigma=5000$



(c) $\sigma=50000$

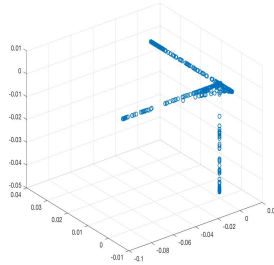
(b) As σ increase, the eigengaps increase about same order as σ does.

(c) The PCA plot represent a frequency space where those pixel points varies the most. I call the space as a frequency space is because that the subspace is a linear transformation from the original frequency space.



(a) PCA

The 3d plot of the first three Laplacian eigenvectors shows that the pixel points can be easily categorized into 3 clusters by kmeans. This 3d space is subspace of graph eigenspace. (knn=15 sigma = 50000)



(a) Eigenspace

Attachment : Codes

0.0.1 Question3a:spectral clustering

```

clear all;
load('SalinasA_corrected.mat');
close all;

data1D = reshape(salinasA_corrected,[],204);

%% Sparse construction
sigma=50;

W_sparse=sparse(size(data1D,1),size(data1D,1));
Knn=15;
NN=zeros(Knn,size(data1D,1));

for i=1:size(data1D,1)
    NN(:,i)=knnsearch(data1D,data1D(i,:), 'k',Knn);
    for j=1:Knn
        W_sparse(i,NN(j,i))=exp(-norm(data1D(i,:)-data1D(NN(j,i),:)).^2/sigma^2);
        % W_sparse(i,NN(j,i))=(norm(data1D(i,:)-data1D(NN(j,i),:)));
    end
end

D_sparse=diag(sum(W_sparse,2));

L_sparse=eye(size(W_sparse))-D_sparse^(-1)*W_sparse;

[EigVecsSparse,EigValsSparse]=eigs(L_sparse,10,'sr');
EigValsSparse=diag(EigValsSparse);

close all;

%% Display Ng, Jordan, Weiss clustering

```

```

Labels=kmeans(EigVecsSparse(:,1:6),6,'Replicates',100);

im = image(reshape(Labels,83,86),'CDataMapping','scaled');
colorbar

%% sigma vs eigenvalues
plot(fliplr(EigValsSparse'))

```

0.0.2 PCA

```

clear all;
load('SalinasA_corrected.mat');
close all;

data1D = reshape(salinasA_corrected,[],204);
pcaM=pca(data1D);
projData=data1D*pcaM;

pca1=reshape(projData(:,1),83,86);

pca1=reshape(projData(:,1),83,86);
image(pca1,'CDataMapping','scaled');

pca2=reshape(projData(:,2),83,86);
image(pca2,'CDataMapping','scaled');

pca3=reshape(projData(:,3),83,86);
image(pca3,'CDataMapping','scaled');

```
