### Question 1

**a.**  $c_r^i=[-\frac{r}{2},\frac{r}{2}]$  is the interval of ith dimension of cube  $C_r^D$ . Because  $Vol_DA=\int_A dx_1...dx_D$ ,

$$Vol_D(C_r^D) = \int_{C_r^D} dx_1 ... dx_D$$

$$= \int_{c_r^D} ... \int_{c_r^1} dx_1 ... dx_D$$

$$= \prod_{i=1}^D \int_{c_r^i} dx_i$$

$$= \prod_{i=1}^D r = r^D$$
(1)

 $\mathbf{b.} \quad \text{Since } A^D_{\epsilon,r} = \{xC^D_r|x \not\in C^D_\epsilon\} = C^D_r - C^D_{r-\epsilon},$ 

$$Vol_D(A_{\epsilon,r}^D) = \int_{A_{\epsilon,r}^D} dx_1...dx_D$$

$$= \int_{C_r^D - C_{r-\epsilon}^D} dx_1...dx_D$$

$$= \int_{C_r^D} dx_1...dx_D - \int_{C_{r-\epsilon}^D} dx_1...dx_D$$

$$= r^D - (r - \epsilon)^D$$
(2)

Then 
$$\frac{Vol_D(A_{\epsilon,r}^D)}{Vol_D(C_r^D)} = \frac{r^D - (r - \epsilon)^D}{r^D} = 1 - (1 - \frac{\epsilon}{r})^D$$

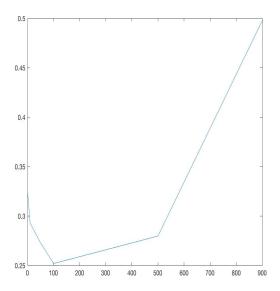
 $\begin{array}{ll} \textbf{c.} & \text{When } D = 10, \text{ if } \epsilon = \frac{r}{10}, \frac{Vol_D(A^D_{\epsilon,r})}{Vol_D(C^D_r)} = 65.13\%. \\ & \text{Because } 0 < \epsilon < r, \ 1 - \frac{\epsilon}{r} < 1 \text{ and } \frac{r}{r - \epsilon} > 1 \end{array}$ 

$$\frac{\partial}{\partial D} \left( \frac{Vol_D(A_{\epsilon,r}^D)}{Vol_D(C_r^D)} \right) = -\frac{\partial}{\partial D} (1 - \frac{\epsilon}{r})^D$$

$$= ln(\frac{r}{r - \epsilon})(1 - \frac{\epsilon}{r}) > 0$$
(3)

It means that the ratio  $\frac{Vol_D(A^D_{\epsilon,r})}{Vol_D(C^D_r)}$  monotonically increases as D get larger.

# Question 2



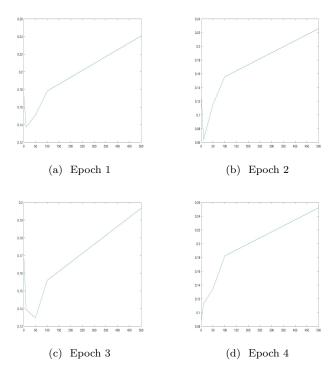
The kNN of best performance among this kNN set,  $\{1, 10, 50, 100, 500, 900\}$ , is at kNN = 100. When kNN is too small, the model will be underfitting; when kNN is too large, the model will be overfitting.

The code will be attached at the end.

#### Question 3

Basing on the results from 10 epoches (below only shows 4 of 10 epoches), when the kNN is smaller 100, the loss will be small. Sometimes when kNN is 1, loss will be relatively large, because a large amount testing points is at boundary of the partitions.

The code will be attached at the end.



## Question 4

a.  $E = \{x \in \mathbb{R}^d | w^Tx = 0\}$ , it means E is nullspace of  $w^T$ ,  $E = Null(w^T)$  Because  $w^T \in \mathbb{R}^d$  and  $w^T \neq 0$ 

$$dim(E) = dim(\mathbb{R}^d) - dim(w^T) = d - 1 \tag{4}$$

**b.** Part (a) has proved that  $S = \{x \in \mathbb{R} | w^T x = 0\}$  has rank d-1. We can rewrite S as below

$$S = \{(x+v) \in \mathbb{R}^d | w^T(x+v) = 0, \text{ for some } v \in \mathbb{R}^d \text{ s.t. } w^Tv = -b\}$$
$$= \{(x+v) \in \mathbb{R}^d | w^Tx = b, \text{ for some } v \in \mathbb{R}^d \text{ s.t. } w^Tv = -b\}$$
 (5)

Now  $E = \{x \in \mathbb{R}^d | w^T x = b\}$  is isomorphic to S by **a translation**, that is

$$E = \{x \in \mathbbm{R}^d | x = y - v \text{ for } y \in S \text{ and some } v \in \mathbbm{R}^d \text{ s.t. } w^T v = -b\}$$

Thus E has same dimension as S, that is d-1.

#### 0.1 Code of Question 2

```
load kNN_ClassifierSyntheticData
% Labels: 1x1000 lebels
% X : 1000X2 Data point
%% Random sample data
XLabels = [X Labels.'];
testidx = randperm(1000,100);
test100 = XLabels(testidx,:) ;
trainidx = setdiff([1:1000],testidx);
%training set
train900 = XLabels(trainidx,:) ;
samX= train900(:,1:2);
samL= train900(:,3);
%initiate data
kNNset = [1 10 50 100 500 900];
Loss100 =[];
%% Model builder
for kNN = kNNset
   Mdl = fitcknn(samX,samL,'NumNeighbors',kNN);
   Loss100 = [Loss100 loss(Mdl,test100(:,1:2),test100(:,3))];
end
load SalinasA_gt
load SalinasA
%% Random sample data
lineSal = reshape(salinasA,[],224);
lineGT = reshape(salinasA_gt,[],1);
XLabels = [lineSal lineGT];
%test set
testidx = randperm(length(XLabels),100);
test100 = XLabels(testidx,:) ;
```

```
%training set
trainidx = setdiff([1:length(XLabels)],testidx);

trainSet = XLabels(trainidx,:);

samX= trainSet(:,1:224);
samL= trainSet(:,225);

%initiate data
kNNset = [1 10 50 100 500];
Loss100 =[];

%% Model builder

for kNN = kNNset
    Mdl = fitcknn(samX,samL,'NumNeighbors',kNN);
    Loss100 = [Loss100 loss(Mdl,test100(:,1:224),test100(:,225))];
end
```