

Gene sorting project

1 Sorting on a two-dimensional lattice

problem statement

Let $\{x[i]\}$ collect points in \mathbb{R}^d . 2D sorting involves a task of placing each point on a node within a lattice. The placement is realized by assigning x to a cortical point $y[n]$ on the lattice such that $d[n] = \min_m d[m]$, $d[m] = \|x - y[m]\|$.

2D sorting is indeed a task of determining all $y[n]$ on a lattice

2 Enumerate nodes on an $M \times M$ lattice

Two index enumeration specifies the joint position of the i th row and j th column on a lattice by (i, j) . Let $n = h(i, j) = M * (i - 1) + j$. h induces one index enumeration of nodes on a lattice.

3 Neighboring relation

Each cortical point $y[n]$ has its neighboring cortical points on a lattice according to contexts of INA gene sorting 38-39. Let $NB(n)$ collect neighboring indices of n on a lattice.

Problem a1: Let Y be a matrix with rows collecting all $y[n]$. Y has $M * M$ rows and d columns. Draw a flow chart to determine the mean of the distance between neighboring cortical points for given Y .

Problem a2: write Matlab codes to implement the flow chart. Assign $y[n]$ to vector $z[n] = ((i-1)/(M-1), (j-1)/(M-1))$, where $n = h(i, j)$. Apply Matlab codes to calculate the mean of the distance between neighboring cortical points and verify the answer.

Gene expressions

Yeast_822 collects 822 yeast gene expressions at seven time courses. Try to sort them on an $M \times M$ lattice, where $M = \text{round}(\sqrt{822})$ and cortical points belong \mathbb{R}^7 .

Problem b1: Apply SOM toolbox to sort yeast822 on a lattice.

Problem b2: Calculate the mean of the distance between neighboring cortical points.

Function approximation

Let $\{y[n]\}$ collect cortical points after sorting yeast822 on a lattice by SOM toolbox.

Problem c1: prepare paired data $S(k) = \{(z[n], y[n, k])\}$ for $k = 1, \dots, d$

Problem c2: learn an RBF network subject to paired data in $S(k)$ for $k = 1, \dots, d$

Problem c3: Let f_k denote the mapping derived by learning an RBF network subject to $S(k)$ and $e(k,t)$ denote the mean square error of fitting f_k to paired data in $S(t)$. Display $e(k,t)$ for all k and t .

Problem c4: Let U collect $e(k,t)$ for $k \leq t$ and L for $k < t$. Check if elements in each row of U increase and those in each row of L decrease strictly.

Classification of SRBCT cancers

There are four types of cancers in the SRBCT database.

Problem d1: For each cancer type, randomly pick five patients and sort their gene expressions on a two-dimensional lattice.

Problem d2: for each cancer c , display approximating functions $f_k(c)$, $k=1\dots 5$, where $c=1\dots 4$

Problem d3: According to gene sorting of each cancer type c on a lattice, gene expressions of a patient form five sets of paired data, denoted by $S(c)$, $c=1\dots 4$. Fitting $f_k(c)$ of different k to $S(c)$ lead to a vector of 20 features.

Problem d4: how to classify four types of SRBCT cancers based on 20 features derived by gene sorting.

Classification of colon cancers

There are two types of patients in the colon database.

Problem e1: For each type, randomly pick five patients and sort their gene expressions on a two-dimensional lattice.

Problem e2: For each type c , display approximating functions $f_k(c)$, $k=1\dots 5$, where $c=1\dots 2$

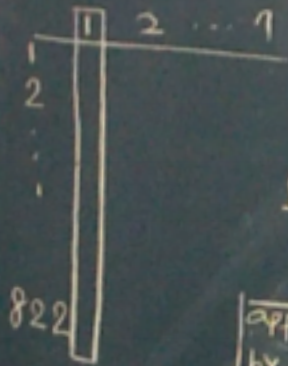
Problem e3: According to gene sorting of each cancer type c on a lattice, gene expressions of a patient form two sets of paired data, denoted by $S(c)$, $c=1,2$. Fitting $f_k(c)$ to $S(c)$ lead to a vector of 10 features.

Problem d4: how to classify two types of patients based on features derived by gene sorting.

Problem f1: refer to INA gene clustering 60

Problem f2: refer to INA gene clustering 61

Yeast₈₂₂ : $Y_{822 \times 7} \Rightarrow$ SOM
2D sorting $\Rightarrow Y_{784 \times 7} : \{y_n \in \mathbb{R}^{77}\}$



RBFN
learning

$$\leftarrow S_k = \{(z_n, y_n[k])\}_n$$

$$f(z; \theta_k)$$

app $y_n(t)$
by $f(z_n; \theta_k)$

$$e_{kt}$$

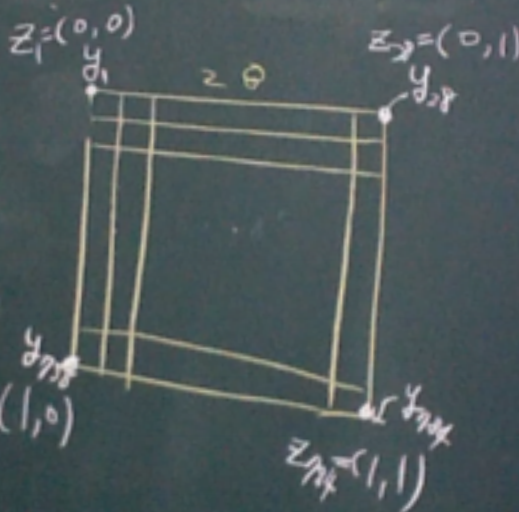
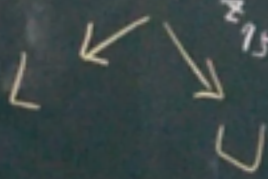
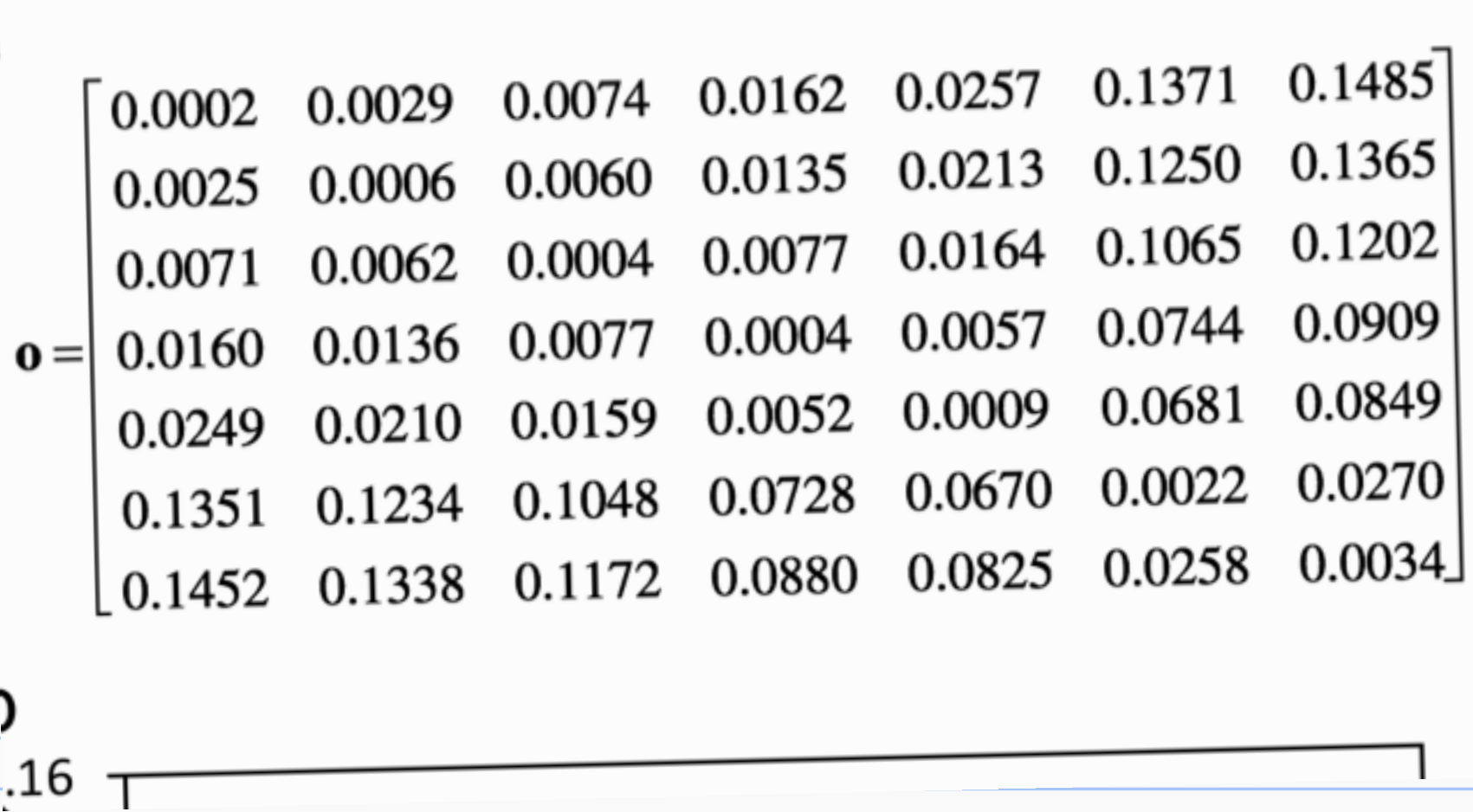


Fig. 10. A plot that collects an SD functional surface

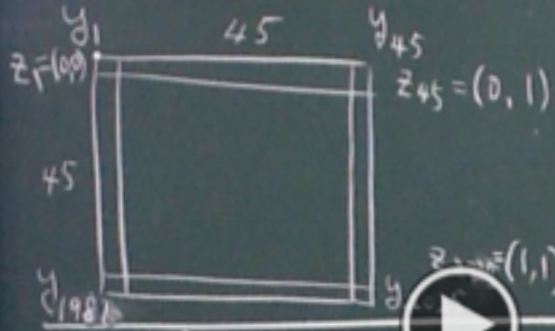


Cancer

$C_{2000 \times 5}$

SOM
Gene sorting

$\Rightarrow Y_{2025 \times 5} \{y_n \in R^5\}$



$S_k = \{(z_n, y_n[k])\}$

RBF learning

$f_1(z; \theta_k)$

normal

$N_{2000 \times 5}$

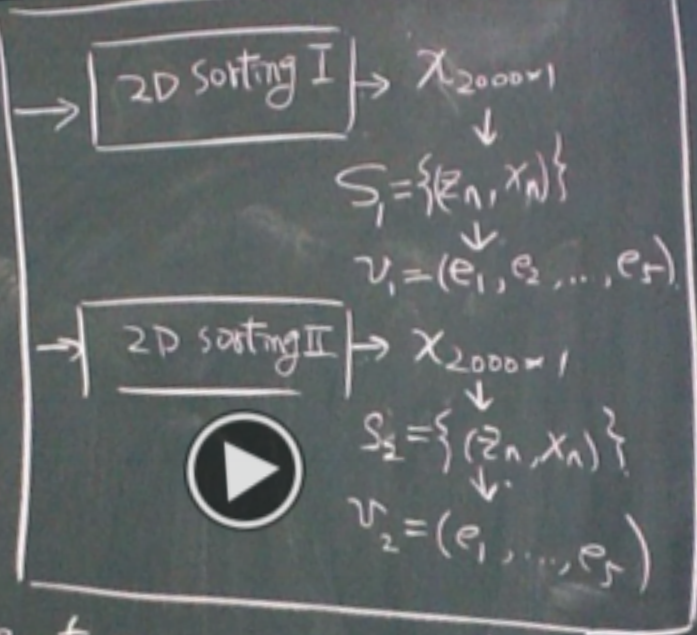
SOM
Gene sorting

$\Rightarrow Y_{2025 \times 5} \{y_n \in R^5\}$

$S_k = \{(z_n, y_n[k])\}$

$f_2(z; \theta_k) \leftarrow$ RBF learning

$X_{2000 \times 1}$ →



e_k sense of app
 x_n by $f(z; \theta_k)$
 $k=1, \dots, 5$

$$V = \begin{bmatrix} v_1 & v_2 \end{bmatrix}$$

$V_{10 \times 63}$

```
function Y=an_kmeans(X,K)
% initialization X: Nxd
beta=0.008; [N d]=size(X);
mx=sum(X,1)/N;
Y = ones(K,1)*mx+rand(K,d)*0.005-0.0025;
hc=0;
while ~hc
    % A calculate cross distance of X and Y
    D=cross_dis(X,Y);
    % B determine overlapping memberships
    % V NxK pp20
    expD=exp(-beta*D);
    sum_expD=sum(expD');
    V=expD./(sum_expD'*ones(1,K));
    % C update Y
    % Y Kxd pp18
    for i=1:K
        Y(i,:)=V(:,i)'+X(:,:);
        if sum(V(:,i).^2) > 1/K*0.01
            Y(i,:)=Y(i,+)/sum(V(:,i));
        else
            Y(i,:)=Y(i,)+rand(1,d)*0.005-0.0025
        end
    end
end
beta = beta/0.9985;
st = mean(sum(V.^2,2));
if st < 1/K+0.005
    Y = Y+rand(K,d)*0.005-0.0025;
end
```

```
iPad 上午11:59 57%
未命名
ftp://admin:512wjm@134.208.26.61/upload/an_kmeans.m Google
while ~hc
    % A calculate cross distance of X and Y
    D=cross_dis(X,Y);
    % B determine overlapping memberships
    % V NxK pp20
    expD=exp(-beta*D);
    sum_expD=sum(expD');
    V=expD./(sum_expD'*ones(1,K));
    % C update Y
    % Y Kxd ppl8
    for i=1:K
        Y(i,:)=V(:,i)*X(:,:);
        if sum(V(:,i).^2) > 1/K*0.01
            Y(i,:)=Y(i,+)/sum(V(:,i));
        else
            Y(i,:)=Y(i,)+rand(1,d)*0.005-0.0025
        end
    end
    beta = beta/0.9985;
    st = mean(sum(V.^2,2));
    if st < 1/K+0.005
        Y = Y+rand(K,d)*0.005-0.0025;
    end
    if st > 0.95
        hc=1;
    end
    fprintf(' %f %f\n',beta, st);
end
```