

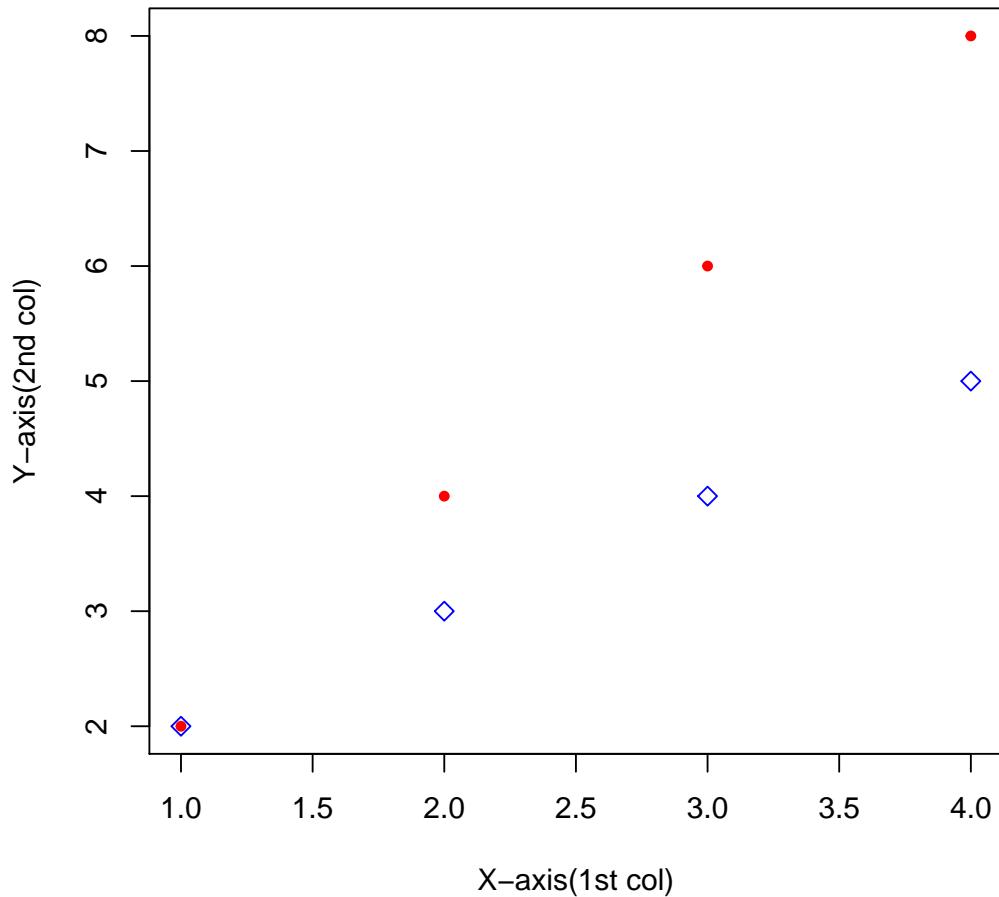
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**Solution for Ex 7.18**

- (a) Use Sweave to accomplish goal, we can get the figure: jhuang63HW3-1.png (uploaded to github file)

**Point Plot of Matrix A and B**



Listing 1: R Codes for Ex 7.18

```
\includegraphics{jhuang63HW3-002}
```

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- (b) The rank of  $A$  is 2 and the rank of  $B$  is 1.

Generating the second and third columns of  $A$  from the first column will change the determinant of this matrix. So, it does not work.

- (c) See the following description:

```

> a = cbind(A[1:2, 3])
> A12 = A[1:2, 1:2]
> dA12 = det(A12)
> x = solve(A12, a)

```

Here is how the linear combination could create the third column from the first two:  
For  $A$ :

$$c3 = -1 * c1 + 2 * c2.$$

For  $B$ :

$$c3 = x * c1 + y * c2$$

for all  $x$  and  $y$  that satisfy  $1 * x + 2 * y = 3$ .

The determinant of the rank 2 submatrix  $A12$  is  $\det(A12) = \det \begin{pmatrix} 1 & 2 \\ 2 & 3 \end{pmatrix} = -1 \neq 0$ ,

So, this matrix can be used with corresponding coordinates in the third column to derive the linear combination.

For  $B$ , the rank is 1, so we can use the first row of it to find the linear combination. The coefficient for the 1st column  $x$  and the 2nd column  $y$  can be any numbers that satisfy following demand:

$$1 * x + 2 * y = 3$$

## Ex 7.24

- [COMMIT] use `lstlisting` to list your R code
- [COMMIT] use R/Sweave for computation, but do not use the built-in `dist` function

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- [COMMIT] use R/Sweave for computation, and this time, do use the built-in `dist` function for comparison

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- [COMMIT] make sure to explain your computation, e.g., compare the two computations

## Solution for Ex 7.24

Listing 2: R codes for Ex 7.24

```
M=A%*%t(A)
c=cbind(diag(M))
one=cbind(c(1,1,1,1))
D=c%*%t(one)+one%*%t(c)-2*M
Dis=sqrt(D)
```

---

- Without Built-in dist function:

```
> M = A %*% t(A)
> c = cbind(diag(M))
> one = cbind(c(1, 1, 1, 1))
> D = c %*% t(one) + one %*% t(c) - 2 * M
> Dis = sqrt(D)
```

- With Built-in dist function:

```
> di = dist(M)
> distance = as.matrix
```

Now, we can calculate the distance matrix:

$$D = \begin{pmatrix} 0 & 3 & 12 & 27 \\ 3 & 0 & 3 & 12 \\ 12 & 3 & 0 & 3 \\ 27 & 12 & 3 & 0 \end{pmatrix}$$

If there are no built-in dist functions can be used, then we shall use the formula in the book to derive a matrix  $M = AA'$ , then calculate the distance matrix from the diagonal entries of  $X$ , matrix  $X$  as well as a one vector.

If there are built-in dist funtions can be used, we just need to use the built-in R DIST function to do the calculation directly.

**Ex 7.30** Omit (c), (d) and (e). The necessary data is saved in `matlabclown.RData` and can be found from the course git folder. The followings are the equivalent R version:

```
load('matlabclown.RData')
image(X) # omit this in your Sweave code
svdX = svd(X)
```

---

```

U = svdX$u
S = diag(svdX$d)
V = svdX$v
k = 10
M = U[,1:k,drop=FALSE] %*% S[1:k,1:k,drop=FALSE] %*% t(V[,1:k,drop=FALSE])
image(M) # omit this in your Sweave code
image(M,col=gray.colors(k))

```

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(a) [COMMIT] choose a small, a medium and a large value for  $k$

- for each  $k$ ,
  - \* do [COMMIT]
  - \* your performance evaluation is to be included as a caption, and change `tinyK`, `width` and `height` accordingly

```

\begin{figure}
\centering
\begin{Schunk}
\begin{Sinput}
> tinyK = 1
\end{Sinput}
\end{Schunk}
\includegraphics{jhuang63HW3-009}
\caption{<YOUR PERFORMANCE EVALUATION> on $1$}
\label{fig:matlabclownKaNumber}
\end{figure}

```

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- (b)
- [COMMIT] code up all your computation using R/Sweave before starting to type your explanation
  - [COMMIT] write your explanation referring to the numbers computed in the previous step, using
-

**Solution for Ex 7.30:**

(b)

- (c) Based on the four figures we have now, it is obvious that clown pictures become clearer as the value of K increasing from small to big. And the corresponding rate are 0.204574122478363,0.574105688285 respectively.

**Problems from Chapter 4: Multidimensional Scaling**

(a) > *tinyK* = 1  
> *smallK* = 20  
> *mediumK* = 80  
> *largeK* = 150  
> *load*("~/Users/hj//550400/jhuang63HW3.git/matlabclown.RData")  
> *svdX* = *svd*(*X*)  
> *U* = *svdX\$u*  
> *S* = *diag*(*svdX\$d*)  
> *V* = *svdX\$v*  
> *k* = *tinyK*  
> *M* = *U*[, 1:*k*, *drop* = FALSE] %\*% *S*[1:*k*, 1:*k*, *drop* = FALSE] %\*%  
+ t(*V*[, 1:*k*, *drop* = FALSE])  
> *image*(*M*, col = gray.colors(*k*))

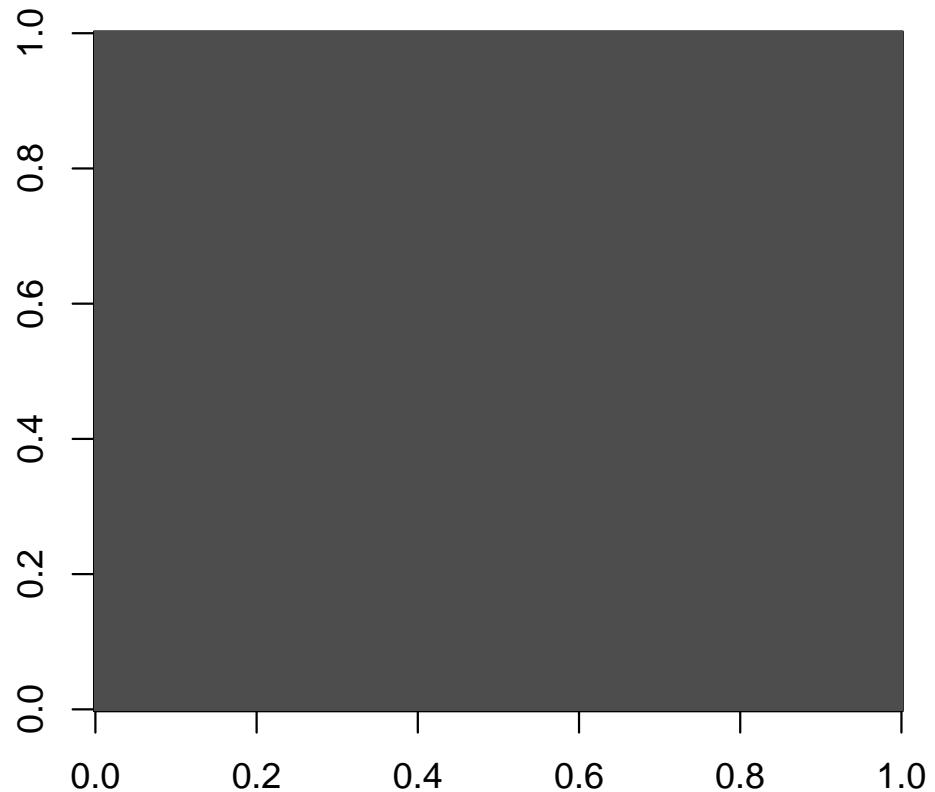


Figure 1: The Poorest Performance on 1

```
> tinyK = 1
> smallK = 20
> mediumK = 80
> largeK = 150
> load("/Users/hj//550400/jhuang63HW3.git/matlabclown.RData")
> svdX = svd(X)
> U = svdX$u
> S = diag(svdX$d)
> V = svdX$v
> k = smallK
> M = U[, 1:k, drop = FALSE] %*% S[1:k, 1:k, drop = FALSE] %*%
+     t(V[, 1:k, drop = FALSE])
> image(M, col = gray.colors(k))
```

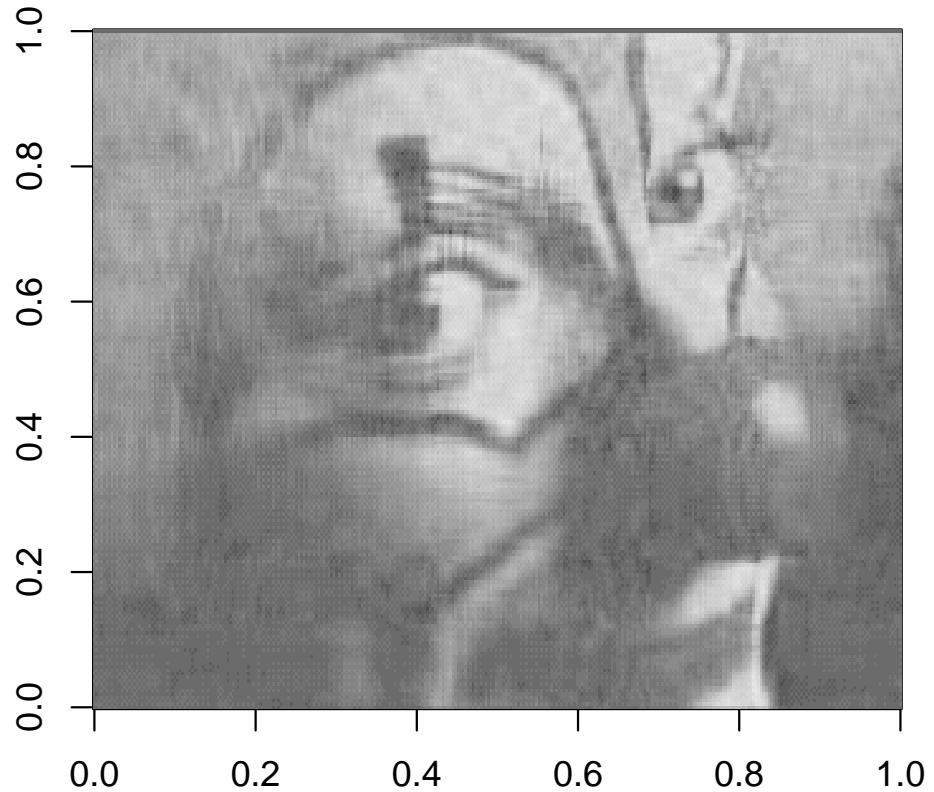


Figure 2: Poor Performance on 20

```
> tinyK = 1
> smallK = 20
> mediumK = 80
> largeK = 150
> load("/Users/hj//550400/jhuang63HW3.git/matlabclown.RData")
> svdX = svd(X)
> U = svdX$u
> S = diag(svdX$d)
> V = svdX$v
> k = mediumK
> M = U[, 1:k, drop = FALSE] %*% S[1:k, 1:k, drop = FALSE] %*%
+     t(V[, 1:k, drop = FALSE])
> image(M, col = gray.colors(k))
```

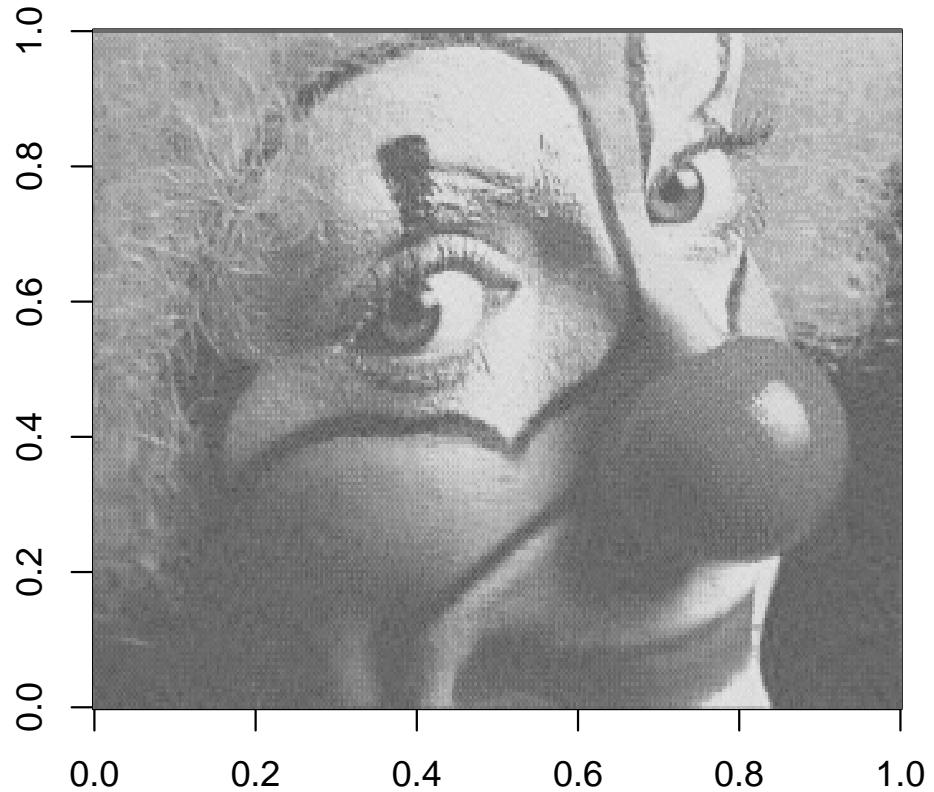


Figure 3: Good Performance on 80

```
> tinyK = 1
> smallK = 20
> mediumK = 80
> largeK = 150
> load("/Users/hj//550400/jhuang63HW3.git/matlabclown.RData")
> svdX = svd(X)
> U = svdX$u
> S = diag(svdX$d)
> V = svdX$v
> k = largeK
> M = U[, 1:k, drop = FALSE] %*% S[1:k, 1:k, drop = FALSE] %*%
+     t(V[, 1:k, drop = FALSE])
> image(M, col = gray.colors(k))
```

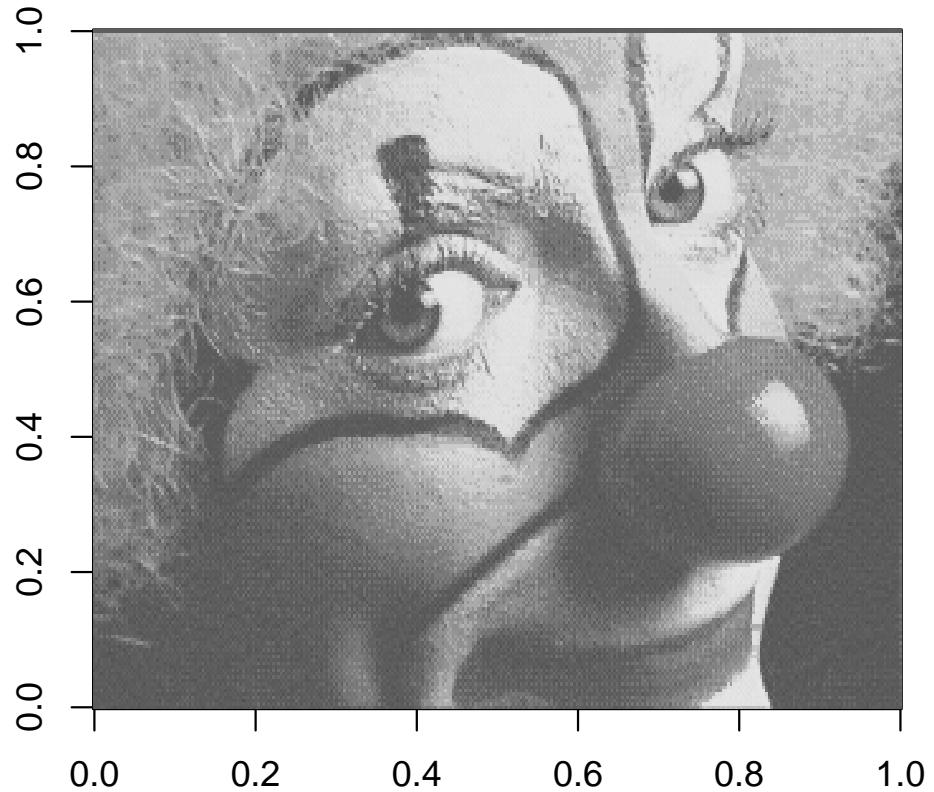


Figure 4: Better Performance on 150

## Solution for Ex 4.1:

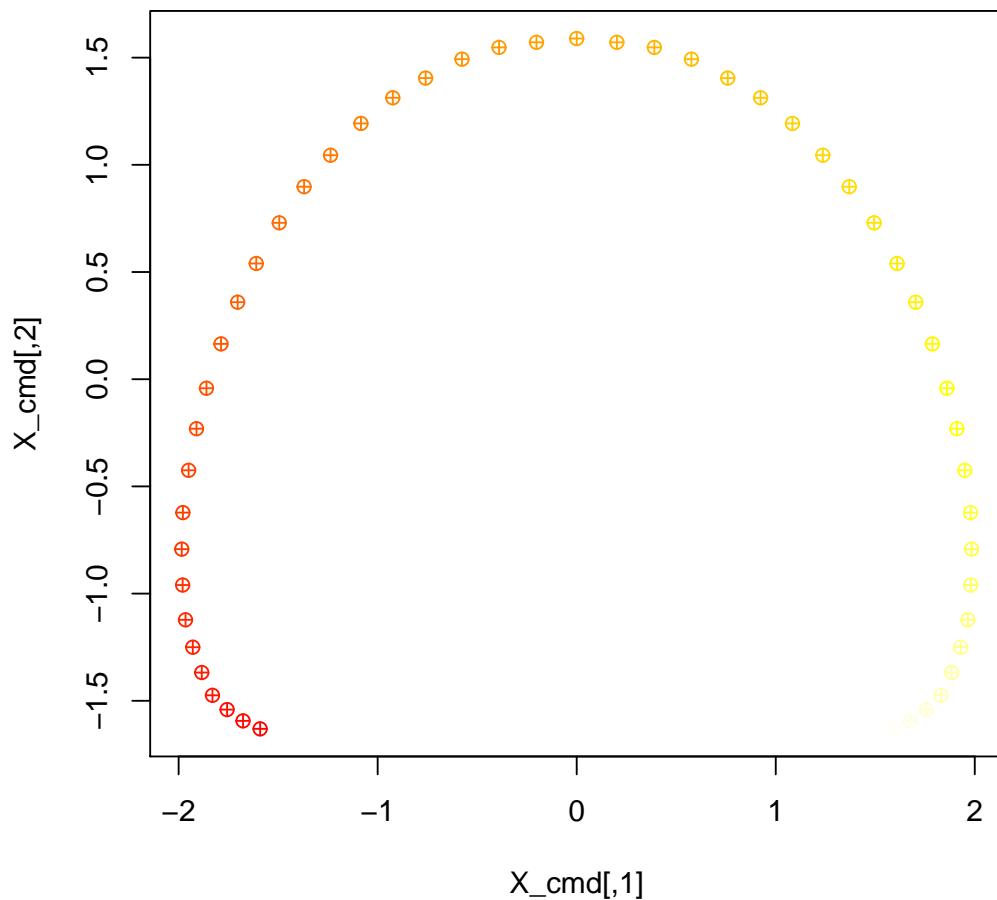
(1, 1)	(2, 1)	(3, 1)	(4, 1)	(5, 1)	(6, 1)	(7, 1)	(8, 1)	(9, 1)	(10, 1)
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Figure 5: first ten objects on a Line

Listing 3: R codes

```
M=matrix(rep(1,51*51),nrow=51,byrow=TRUE)
for (i in 1:51){
  for (j in 1:51){
    if(i == j) M[i,j]=9
    else if(abs(i-j)>=22 & abs(i-j)<=24) M[i,j]=1
    else if(abs(i-j)>=19 & abs(i-j)<=21) M[i,j]=2
    else if(abs(i-j)>=16 & abs(i-j)<=18) M[i,j]=3
    else if(abs(i-j)>=13 & abs(i-j)<=15) M[i,j]=4
    else if(abs(i-j)>=10 & abs(i-j)<=12) M[i,j]=5
    else if(abs(i-j)>=7 & abs(i-j)<=9) M[i,j]=6
    else if(abs(i-j)>=4 & abs(i-j)<=6) M[i,j]=7
    else if(abs(i-j)>=1 & abs(i-j)<=3) M[i,j]=8
    else M[i,j]=0
  }
}
D=matrix(rep(1,51*51),nrow=51,byrow=TRUE)
for (i in 1:51){
  for (j in 1:51){
    D[i,j]=sqrt(M[i,i]+M[j,j]-2*M[i,j])
  }
}
X_cmd=cmdscale(D)
plot(X_cmd,col=heat.colors(51),pch=10)
```

---



From the graphic above, every point is similar to its neighbors while the similarity dies down with distance. The dissimilarity in color represents the dissimilarity of each point.

### Ex 4.3

- [COMMIT] list your R code using `lstlisting`
- [COMMIT] load the data (`require(MVA); data(gardenflowers)`) and compute using R/Sweave
- [COMMIT] include a plot of (relative) positions using R/Sweave
- [COMMIT] allocate at least a quarter page of *text* explaining the result

#### Solution for Ex 4.3:

Listing 4: R codes for Flowers Comparison

```
require(MVA)
data(gardenflowers)
flowers_m=cmdscale(gardenflowers,k=17,eig=TRUE)
x=flowers_m$points[,1]
```

```

y=flowers_m$points[,2]
plot(x,y,xlab = "First column of input", ylab = "Second column of input", xlim = range(x)*1.5, type = "n")
text(x, y, labels = colnames(as.matrix(gardenflowers)), cex = 0.5)

```

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