

DSA 8070 R Session 13: Multidimensional Scaling

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Classical Multidimensional Scaling

Distances Between US Cities

```
loc <- cmdscale(UScitiesD)
x <- loc[, 1]; y <- loc[, 2]
plot(x, y, type = "n", xlab = "", ylab = "", asp = 1,
     axes = FALSE, main = "cmdscale(UScitiesD)")
text(x, y, rownames(loc), cex = 0.8)
```

cmdscale(UScitiesD)



```
# Flip Axes
x1 <- -loc[, 1]; y1 <- -loc[, 2]
plot(x1, y1, type = "n", xlab = "", ylab = "", asp = 1,
```

```
axes = FALSE, main = "cmdscale(UScitiesD)")
text(x1, y1, rownames(loc), cex = 0.8)
```

cmdscale(UScitiesD)



Air Pollution in US Cities

```
library(HSAUR3)
```

```
## Loading required package: tools
```

```
data(USairpollution)
dat <- USairpollution
head(dat)
```

```
##           S02 temp manu popul wind precip predays
## Albany      46 47.6  44  116  8.8  33.36    135
## Albuquerque 11 56.8  46  244  8.9   7.77     58
## Atlanta     24 61.5 368  497  9.1  48.34    115
## Baltimore   47 55.0 625  905  9.6  41.31    111
## Buffalo     11 47.1 391  463 12.4  36.11    166
## Charleston  31 55.2  35   71  6.5  40.75    148
```

```
summary(dat)
```

```
##           S02           temp           manu           popul
## Min.      : 8.00    Min.    :43.50    Min.      : 35.0    Min.      : 71.0
## 1st Qu.: 13.00    1st Qu.:50.60    1st Qu.: 181.0    1st Qu.: 299.0
## Median : 26.00    Median :54.60    Median : 347.0    Median : 515.0
## Mean     : 30.05    Mean     :55.76    Mean      :463.1    Mean      :608.6
## 3rd Qu.: 35.00    3rd Qu.:59.30    3rd Qu.: 462.0    3rd Qu.: 717.0
## Max.     :110.00    Max.      :75.50    Max.     :3344.0    Max.     :3369.0
##           wind           precip           predays
## Min.      : 6.000    Min.      : 7.05    Min.      : 36.0
## 1st Qu.:  8.700    1st Qu.:30.96    1st Qu.:103.0
## Median :  9.300    Median :38.74    Median :115.0
```

```
## Mean   : 9.444   Mean   :36.77   Mean   :113.9
## 3rd Qu.:10.600   3rd Qu.:43.11   3rd Qu.:128.0
## Max.   :12.700   Max.    :59.80   Max.    :166.0
```

```
xs <- apply(dat, 2, function(x) (x - min(x)) / (diff(range(x))))
summary(xs)
```

```
##          SO2              temp              manu              popul
## Min.   :0.00000   Min.   :0.00000   Min.   :0.00000   Min.   :0.00000
## 1st Qu.:0.04902   1st Qu.:0.2219   1st Qu.:0.04412   1st Qu.:0.06913
## Median :0.17647   Median :0.3469   Median :0.09429   Median :0.13463
## Mean   :0.21616   Mean   :0.3832   Mean   :0.12937   Mean   :0.16301
## 3rd Qu.:0.26471   3rd Qu.:0.4938   3rd Qu.:0.12904   3rd Qu.:0.19588
## Max.   :1.00000   Max.    :1.0000   Max.    :1.00000   Max.    :1.00000
##          wind          precip          predays
## Min.   :0.0000   Min.   :0.0000   Min.   :0.0000
## 1st Qu.:0.4030   1st Qu.:0.4533   1st Qu.:0.5154
## Median :0.4925   Median :0.6008   Median :0.6077
## Mean   :0.5140   Mean   :0.5634   Mean   :0.5992
## 3rd Qu.:0.6866   3rd Qu.:0.6836   3rd Qu.:0.7077
## Max.   :1.0000   Max.    :1.0000   Max.    :1.0000
```

```
## compute distance matrix
poldist <- dist(xs)
## reduce to 2 dimensions
(pol.mds <- cmdscale(poldist, k = 2, eig = TRUE))
```

```
## $points
##          [,1]      [,2]
## Albany      0.140558172 -0.046859954
## Albuquerque -0.364824787 -0.636602091
## Atlanta     -0.155922591  0.244511276
## Baltimore    0.153189990  0.067519907
## Buffalo     0.256244063  0.003022604
## Charleston  -0.128730958  0.215783429
## Chicago      1.197000315  0.009638168
## Cincinnati  -0.084166097  0.106828800
## Cleveland    0.531787447  0.056305378
## Columbus     0.025412911  0.033574934
## Dallas      -0.258008194 -0.062640448
## Denver       -0.110682033 -0.510378502
## Des Moines   -0.007603614 -0.244344703
## Detroit      0.341537781 -0.105917971
## Hartford     0.206766531  0.105259858
## Houston      -0.188167760  0.243707765
## Indianapolis  0.069589745  0.010565926
## Jacksonville -0.349520267  0.412490203
## Kansas City  -0.106424371 -0.085118726
## Little Rock  -0.355970056  0.194004542
## Louisville   -0.046780470  0.144850917
## Memphis      -0.249259311  0.208737990
## Miami        -0.449823739  0.604996816
## Milwaukee    0.217298744 -0.249612250
## Minneapolis  0.326439578 -0.242858309
## Nashville    -0.215002650  0.211835269
```

```

## New Orleans      -0.410715158  0.438263300
## Norfolk          -0.066285208  0.149134571
## Omaha            -0.063335982 -0.241936316
## Philadelphia      0.521031706  0.081089446
## Phoenix          -0.695773353 -0.527859295
## Pittsburgh        0.314965899  0.074640031
## Providence        0.466505620  0.110503750
## Richmond         -0.191967563  0.140461889
## Salt Lake City   -0.111111665 -0.461383196
## San Francisco    -0.253430076 -0.401897024
## Seattle           0.170829143  0.147411289
## St. Louis         0.162208664 -0.016576959
## Washington       -0.031338057  0.041417952
## Wichita          -0.149744969 -0.268806546
## Wilmington       -0.056777379  0.046236280
##
## $eig
## [1]  4.456648e+00  2.819944e+00  2.256196e+00  1.651762e+00  6.199354e-01
## [6]  1.904906e-01  3.068220e-02  1.558353e-15  9.406328e-16  2.494225e-16
## [11] 1.736021e-16  1.471280e-16  1.356518e-16  8.017147e-17  7.511957e-17
## [16] 6.686099e-17  5.684599e-17  5.034791e-17  4.025565e-17  3.312471e-17
## [21] 2.974204e-17  1.555983e-17  1.132251e-17  3.668800e-18 -5.206488e-18
## [26] -8.948794e-18 -9.519928e-18 -1.506805e-17 -1.853275e-17 -2.314710e-17
## [31] -2.858271e-17 -3.093804e-17 -3.151435e-17 -3.396470e-17 -7.209856e-17
## [36] -7.714641e-17 -1.524915e-16 -2.390840e-16 -2.833661e-16 -3.238640e-16
## [41] -1.263609e-15
##
## $x
## NULL
##
## $ac
## [1] 0
##
## $GOF
## [1] 0.6050889 0.6050889
## reduce to 3 dimensions
(pol.mds3 <- cmdscale(poldist, k = 3, eig = TRUE))

```

```

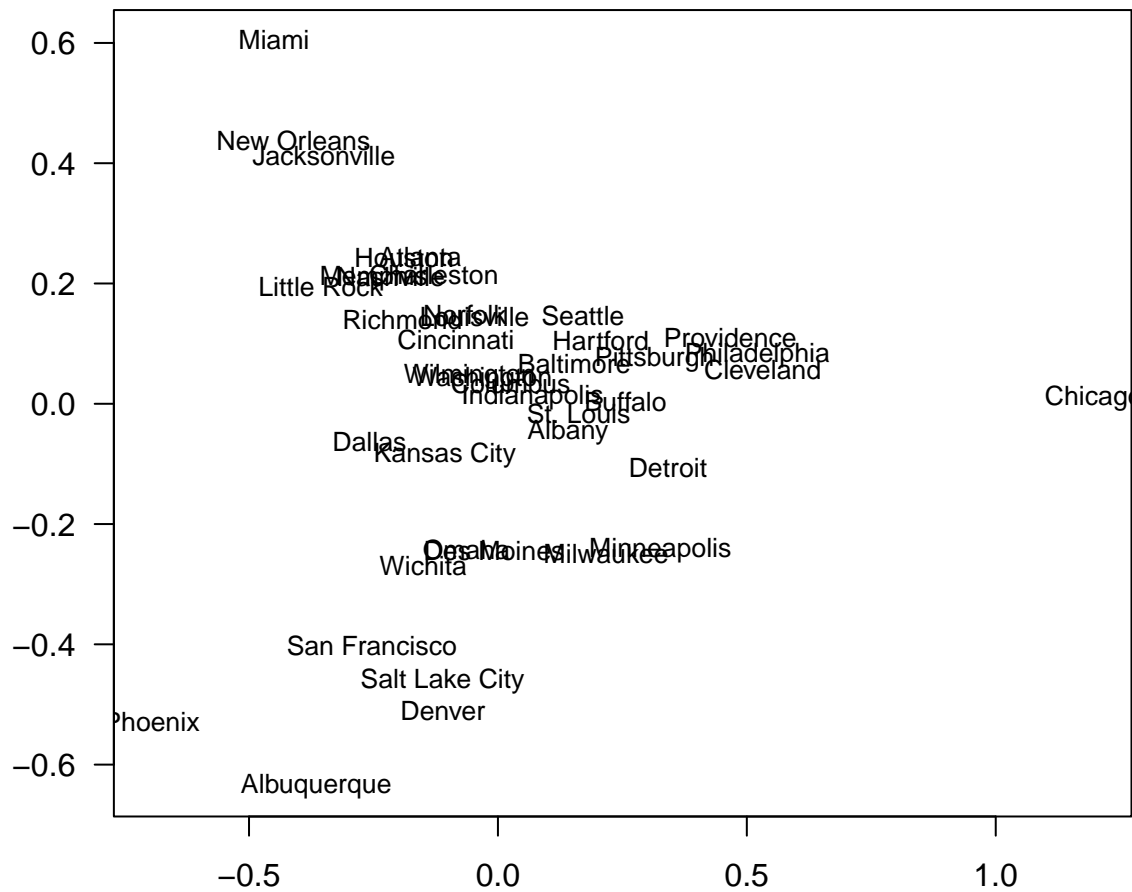
## $points
##           [,1]      [,2]      [,3]
## Albany      0.140558172 -0.046859954  0.267632311
## Albuquerque -0.364824787 -0.636602091 -0.102087912
## Atlanta     -0.155922591  0.244511276 -0.046494117
## Baltimore    0.153189990  0.067519907 -0.089541686
## Buffalo      0.256244063  0.003022604  0.495974986
## Charleston  -0.128730958  0.215783429  0.150107702
## Chicago      1.197000315  0.009638168 -0.824422653
## Cincinnati  -0.084166097  0.106828800  0.039366111
## Cleveland    0.531787447  0.056305378  0.120608309
## Columbus     0.025412911  0.033574934  0.147295326
## Dallas       -0.258008194 -0.062640448 -0.244388882
## Denver       -0.110682033 -0.510378502 -0.029698665
## Des Moines   -0.007603614 -0.244344703  0.268379415
## Detroit      0.341537781 -0.105917971 -0.064362771

```

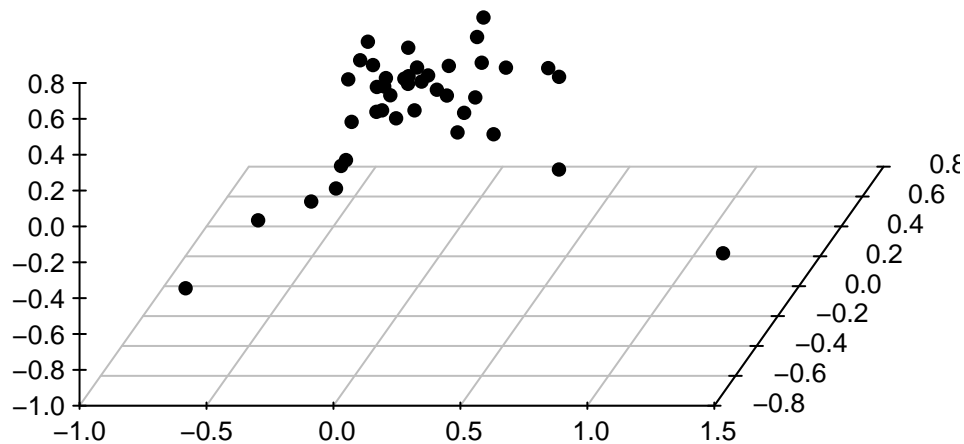
```

## Hartford      0.206766531  0.105259858  0.158528863
## Houston      -0.188167760  0.243707765 -0.266857401
## Indianapolis   0.069589745  0.010565926  0.086925283
## Jacksonville  -0.349520267  0.412490203 -0.110633886
## Kansas City   -0.106424371 -0.085118726  0.051579961
## Little Rock   -0.355970056  0.194004542 -0.008211894
## Louisville    -0.046780470  0.144850917  0.020360112
## Memphis       -0.249259311  0.208737990 -0.062727649
## Miami         -0.449823739  0.604996816 -0.140722158
## Milwaukee     0.217298744 -0.249612250  0.271313047
## Minneapolis   0.326439578 -0.242858309  0.255203375
## Nashville     -0.215002650  0.211835269 -0.015978258
## New Orleans   -0.410715158  0.438263300 -0.105127995
## Norfolk       -0.066285208  0.149134571  0.095300814
## Omaha         -0.063335982 -0.241936316  0.174099908
## Philadelphia   0.521031706  0.081089446 -0.416784375
## Phoenix       -0.695773353 -0.527859295 -0.571707666
## Pittsburgh     0.314965899  0.074640031  0.156669771
## Providence     0.466505620  0.110503750  0.124107741
## Richmond      -0.191967563  0.140461889 -0.002992969
## Salt Lake City -0.111111665 -0.461383196  0.055129138
## San Francisco -0.253430076 -0.401897024 -0.193036023
## Seattle        0.170829143  0.147411289  0.266888111
## St. Louis      0.162208664 -0.016576959 -0.128585601
## Washington    -0.031338057  0.041417952 -0.053942965
## Wichita       -0.149744969 -0.268806546  0.140296711
## Wilmington    -0.056777379  0.046236280  0.132538529
##
## $eig
## [1]  4.456648e+00  2.819944e+00  2.256196e+00  1.651762e+00  6.199354e-01
## [6]  1.904906e-01  3.068220e-02  1.558353e-15  9.406328e-16  2.494225e-16
## [11] 1.736021e-16  1.471280e-16  1.356518e-16  8.017147e-17  7.511957e-17
## [16] 6.686099e-17  5.684599e-17  5.034791e-17  4.025565e-17  3.312471e-17
## [21] 2.974204e-17  1.555983e-17  1.132251e-17  3.668800e-18 -5.206488e-18
## [26] -8.948794e-18 -9.519928e-18 -1.506805e-17 -1.853275e-17 -2.314710e-17
## [31] -2.858271e-17 -3.093804e-17 -3.151435e-17 -3.396470e-17 -7.209856e-17
## [36] -7.714641e-17 -1.524915e-16 -2.390840e-16 -2.833661e-16 -3.238640e-16
## [41] -1.263609e-15
##
## $x
## NULL
##
## $ac
## [1] 0
##
## $GOF
## [1] 0.792704 0.792704
## plot
par(las = 1, mgp = c(2, 1, 0), mar = c(3, 3, 1, 0.5))
x <- pol.mds$points
plot(x[, 1], x[, 2], type = "n", xlab = "", ylab = "")
text(x[, 1], x[, 2], labels = rownames(x), cex = 0.8)

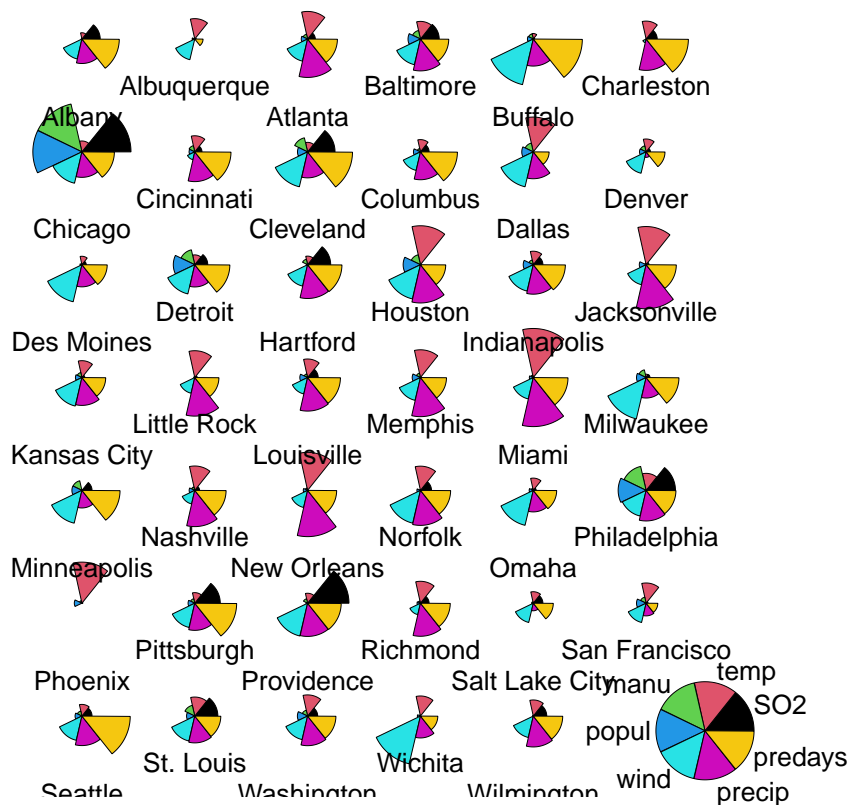
```



```
library(scatterplot3d)
par(las = 1)
scatterplot3d(pol.mds3$points, pch = 16, angle = 75,
              box = F, xlab = "", ylab = "",
              zlab = "")
```



```
stars(xs, draw.segments = TRUE, key.loc = c(15, 2))
```



Non-Metric Multidimensional Scaling

House of Representatives Voting Data

```
data("voting", package = "HSAUR2")
voting
```

##	Hunt(R)	Sandman(R)	Howard(D)	Thompson(D)	Freylinghuysen(R)	
## Hunt(R)	0	8	15	15	10	
## Sandman(R)	8	0	17	12	13	
## Howard(D)	15	17	0	9	16	
## Thompson(D)	15	12	9	0	14	
## Freylinghuysen(R)	10	13	16	14	0	
## Forsythe(R)	9	13	12	12	8	
## Widnall(R)	7	12	15	13	9	
## Roe(D)	15	16	5	10	13	
## Heltoski(D)	16	17	5	8	14	
## Rodino(D)	14	15	6	8	12	
## Minish(D)	15	16	5	8	12	
## Rinaldo(R)	16	17	4	6	12	
## Maraziti(R)	7	13	11	15	10	
## Daniels(D)	11	12	10	10	11	
## Patten(D)	13	16	7	7	11	
##	Forsythe(R)	Widnall(R)	Roe(D)	Heltoski(D)	Rodino(D)	Minish(D)
## Hunt(R)	9	7	15	16	14	15
## Sandman(R)	13	12	16	17	15	16
## Howard(D)	12	15	5	5	6	5

## Thompson(D)	12	13	10	8	8	8
## Freylinghuysen(R)	8	9	13	14	12	12
## Forsythe(R)	0	7	12	11	10	9
## Widnall(R)	7	0	17	16	15	14
## Roe(D)	12	17	0	4	5	5
## Heltoski(D)	11	16	4	0	3	2
## Rodino(D)	10	15	5	3	0	1
## Minish(D)	9	14	5	2	1	0
## Rinaldo(R)	10	15	3	1	2	1
## Maraziti(R)	6	10	12	13	11	12
## Daniels(D)	6	11	7	7	4	5
## Patten(D)	10	13	6	5	6	5
##						
	Rinaldo(R)	Maraziti(R)	Daniels(D)	Patten(D)		
## Hunt(R)	16	7	11	13		
## Sandman(R)	17	13	12	16		
## Howard(D)	4	11	10	7		
## Thompson(D)	6	15	10	7		
## Freylinghuysen(R)	12	10	11	11		
## Forsythe(R)	10	6	6	10		
## Widnall(R)	15	10	11	13		
## Roe(D)	3	12	7	6		
## Heltoski(D)	1	13	7	5		
## Rodino(D)	2	11	4	6		
## Minish(D)	1	12	5	5		
## Rinaldo(R)	0	12	6	4		
## Maraziti(R)	12	0	9	13		
## Daniels(D)	6	9	0	9		
## Patten(D)	4	13	9	0		

```
library(MASS)
voting_mds <- isoMDS(voting, k = 2)
```

```
## initial value 15.268246
## iter 5 value 10.264075
## final value 9.879047
## converged
```

```
str(voting_mds)
```

```
## List of 2
## $ points: num [1:15, 1:2] -8.44 -7.41 6.09 3.52 -7.25 ...
## .. attr(*, "dimnames")=List of 2
## .. ..$ : chr [1:15] "Hunt(R)" "Sandman(R)" "Howard(D)" "Thompson(D)" ...
## .. ..$ : NULL
## $ stress: num 9.88
```

```
par(las = 1)
plot(voting_mds$points, type = "n", xlim = c(-15, 10),
     xlab = "Coordinate 1", ylab = "Coordinate 2")
text(voting_mds$points, labels = rownames(voting_mds$points))
```