

02_q9

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```
library(stats)
city_distances <- matrix(c(
  0, 587, 1212, 701, 1936, 604, 748, 2139, 2182, 543,
  587, 0, 920, 940, 1745, 1188, 713, 1858, 1737, 597,
  1212, 920, 0, 879, 831, 1726, 1611, 1949, 2204, 1494,
  701, 940, 879, 0, 1374, 968, 1420, 1645, 1891, 1220,
  1936, 1745, 831, 1374, 0, 2339, 2451, 347, 2734, 2300,
  604, 1188, 1726, 968, 2339, 0, 1092, 2594, 2408, 923,
  748, 713, 1611, 1420, 2451, 1092, 0, 2571, 678, 205,
  2139, 1858, 1949, 1645, 347, 2594, 2571, 0, 678, 2442,
  2182, 1737, 2204, 1891, 2734, 2408, 678, 678, 0, 2329,
  543, 597, 1494, 1220, 2300, 923, 205, 2442, 2329, 0
), nrow = 10, byrow = TRUE)

# Assigning names to row and columns
city_names <- c("Atlanta", "Chicago", "Denver", "Houston", "Los Angeles", "Miami",
               "New York", "San Francisco", "Seattle", "Washington D.C.")
rownames(city_distances) <- city_names
colnames(city_distances) <- city_names

## A)
## Get dissimilarity distance as city.dissimilarity object
city.dissimilarity <- as.dist(city_distances)

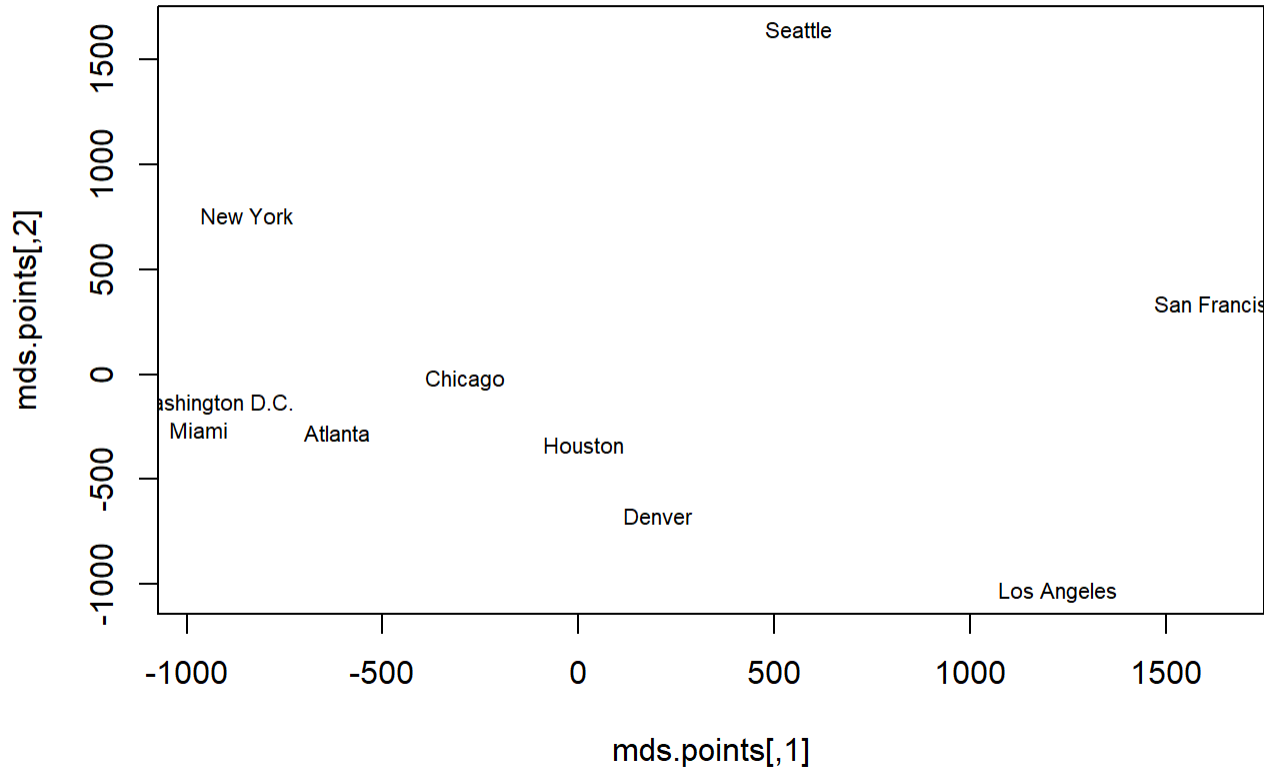
## B)
## Fit the classical MDS model using city.dissimilarity object
mds.model <- cmdscale(city.dissimilarity, eig = TRUE, k = 2) # Dimension 2

## C)
# Summary of model
mds.points <- mds.model$points
print(mds.points)
```

##	[,1]	[,2]
## Atlanta	-616.46326	-277.03319
## Chicago	-288.61063	-22.16151
## Denver	202.61148	-672.61019
## Houston	14.25242	-335.54496
## Los Angeles	1225.78174	-1033.78934
## Miami	-968.45797	-264.31832
## New York	-845.50822	757.66327
## San Francisco	1645.58380	339.92746
## Seattle	563.12009	1646.43854
## Washington D.C.	-932.30945	-138.57175

```
## Interpretation
#

## D)
## Bi-plot of the model
plot(mds.points, type = "n")
text(mds.points, labels = city_names, cex = 0.7)
```



```
## Interpretation
```