

Bioinformatics and Statistical Genetics: Population Substructure

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```
library(genetics)
library(data.table)
library(MASS)
```

Task1

The file SNPChr20.rda contains genotype information of 310 individuals of unknown background. The genotype information concerns 50.000 SNPs on chromosome 20. Load this data into the R environment. The data file contains a matrix Y containing the allele counts (0,1 or 2) for 50.000 SNPs for one of the alleles of each SNP

```
chunk_size <- 10000
raw_data <- fread(file="Chr21.dat", sep = " ", header = TRUE, nThread = chunk_size)
raw_data_df <- data.frame(raw_data)
```

1

1. How many variants are there in this database? What percentage of the data is missing?

```
variants_df <- raw_data_df[,7:ncol(raw_data_df)]

# Convert values different from 0, 1, or 2 to NA
variants_df[!sapply(variants_df, function(x) x %in% c(0, 1, 2))] <- NA

n <- nrow(variants_df) # individuals
p <- ncol(variants_df) # SNPs
cat("1. How many variants are there in this database? \n")

## 1. How many variants are there in this database?

cat(p)

## 138106

cat("\n")

cat("1. What percentage of the data is missing? \n")

## 1. What percentage of the data is missing?

mis <- 100*sum(is.na(variants_df))/(n*p)
cat(mis)

## 0
```

2

2. Compute the Manhattan distance matrix between the individuals (which is identical to

the Minkowsky distance with parameter $\lambda = 1$) using R function `dist`. Include a submatrix of dimension 5 by 5 with the distances between the first 5 individuals in your report

```
distances <- dist(variants_df, method="manhattan", diag = FALSE)
D <- as.matrix(distances)

print("Manhattan distance matrix between first 5 individuals: ")
## [1] "Manhattan distance matrix between first 5 individuals: "
D[1:5, 1:5]
```

##	1	2	3	4	5
## 1	0	53495	55007	58174	53794
## 2	53495	0	55372	55995	55699
## 3	55007	55372	0	54815	55683
## 4	58174	55995	54815	0	59046
## 5	53794	55699	55683	59046	0

3

3. (1p) How does the Manhattan distance relate to the allele sharing distance?

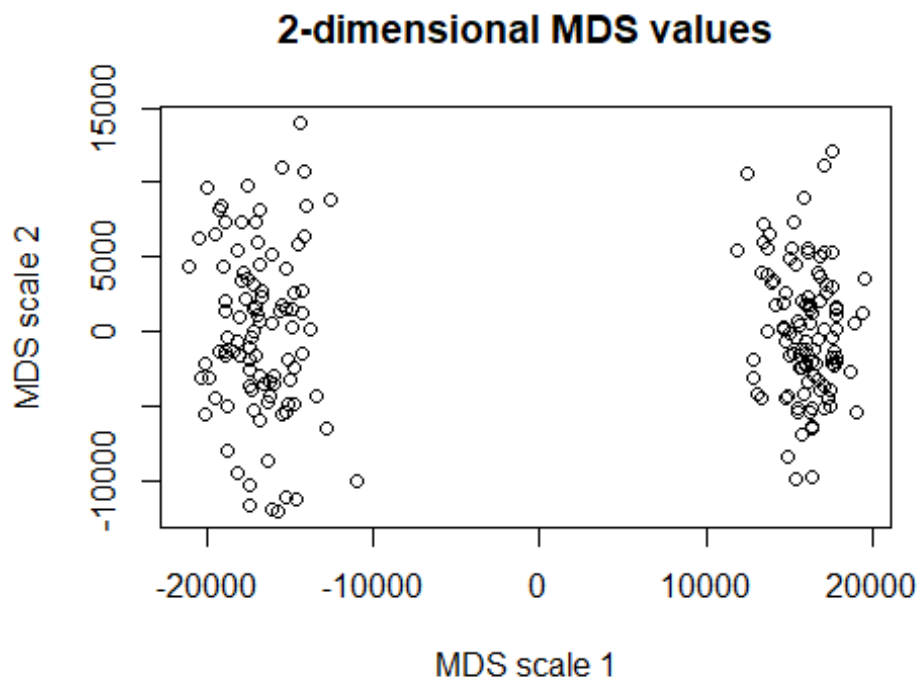
When data is coded in (0,1,2) format, the Manhattan distance is equivalent to allele sharing distance. In particular, the Manhattan distance is directly proportional to the allele sharing distance.

4

4. Apply metric multidimensional scaling (`cmdscale`) with two dimensions, $k = 2$, using the Manhattan distance matrix and include the map in your report. Do you think the data come from one homogeneous human population? If not, how many subpopulations do you think the data might come from, and how many individuals pertain to each subpopulation?

```
mds <- cmdscale(D, k=2, eig=TRUE)
points <- mds$points[,1:2]

plot(points[, 1], points[, 2], main="2-dimensional MDS values", xlab="MDS scale 1",
      ylab="MDS scale 2")
```



```
cat("The map shows that there are 2 different population groups. Thus, the data does not come from a homogenous population.")
```

```
## The map shows that there are 2 different population groups. Thus, the data does not come from a homogenous population.
```

```
subpopulation_1 <- points[points[, 1] > 10000]
```

```
subpopulation_2 <- points[points[, 1] < 10000]
```

```
cat("In the first subpopulation there are ", length(subpopulation_1), "individuals.\n")
```

```
## In the first subpopulation there are 208 individuals.
```

```
cat("In the second subpopulation there are ", length(subpopulation_2), "individuals.")
```

```
## In the second subpopulation there are 198 individuals.
```

5

**5.What is the goodness-of-fit of the two-dimensional approximation to your distance matrix?
Explain which criterium you have used.**

Since we are in the context of Metric MDS, the criterium used to calculate the goodness-of-fit of the k-dimensional approximation of the distance matrix is given:

$(\sum_{j=1}^k (\lambda_j)) / (\sum_{j=1}^n (T_i(\lambda_j)))$, where λ_j are the eigenvalues (sorted in decreasing order) of the matrix, and $T_1(v)=|v|$, and $T2(v)=\max(v,0)$.

where n is the original dimension. In our case $k=2$.

```
gof <- mds$GOF
gof
## [1] 0.1703581 0.1703605
```

6

6. Make a plot of the estimated distances (according to your map of individuals) versus the observed distances. What do you observe? Regress estimated distances on observed distances and report the coefficient of determination of the regression (you can use the function `lm`).

The R-squared value of 0.84284 indicates that approximately 84.28% of the variability in the observed distances can be explained by the estimated distances using the linear relationship shown in the plot. This suggests a positive linear relationship between the estimated and observed distances. It is worth noting that the estimated distances are on a smaller scale than the observed distances. This may suggest that the approximation reduces the distances.

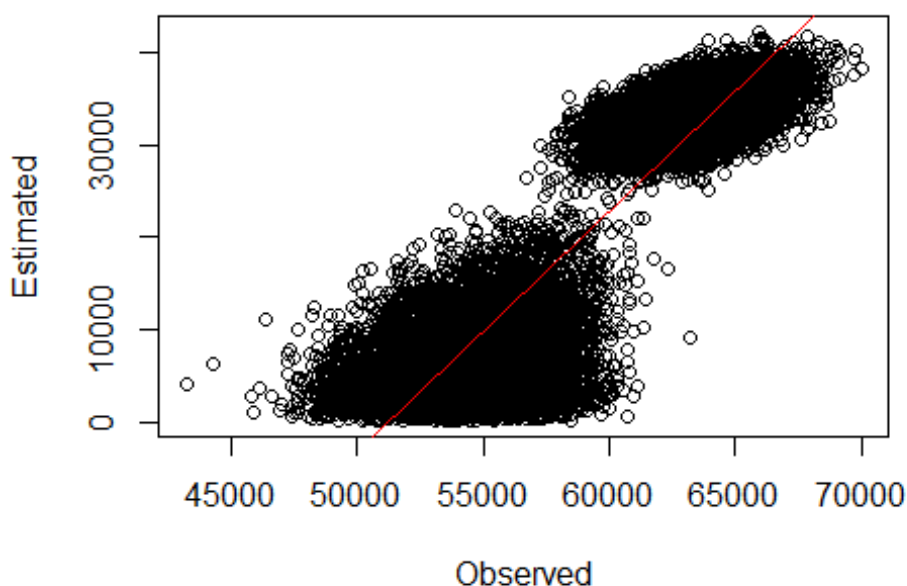
```
estimated_distances <- as.matrix(dist(points))
estimated_distances <- estimated_distances[lower.tri(estimated_distances)]
observed_distances <- D[lower.tri(D)]

plot(observed_distances, estimated_distances, xlab="Observed", ylab="Estimated", main="Estimated vs Observed distances")

regression_model <- lm( estimated_distances ~ observed_distances)

abline(regression_model, col = "red")
```

Estimated vs Observed distances



```
# coefficient of determination (R-squared)
cat("Coefficient of Determination (R-squared): ", summary(regression_model)$r.squa
red, "\n")

## Coefficient of Determination (R-squared): 0.8428413
```

7

7. We now try a (two-dimensional) non-metric multidimensional scaling using the isoMDS function that you will find in MASS library. We use a random initial configuration and, for the sake of reproducibility, make this random initial configuration with the instructions:

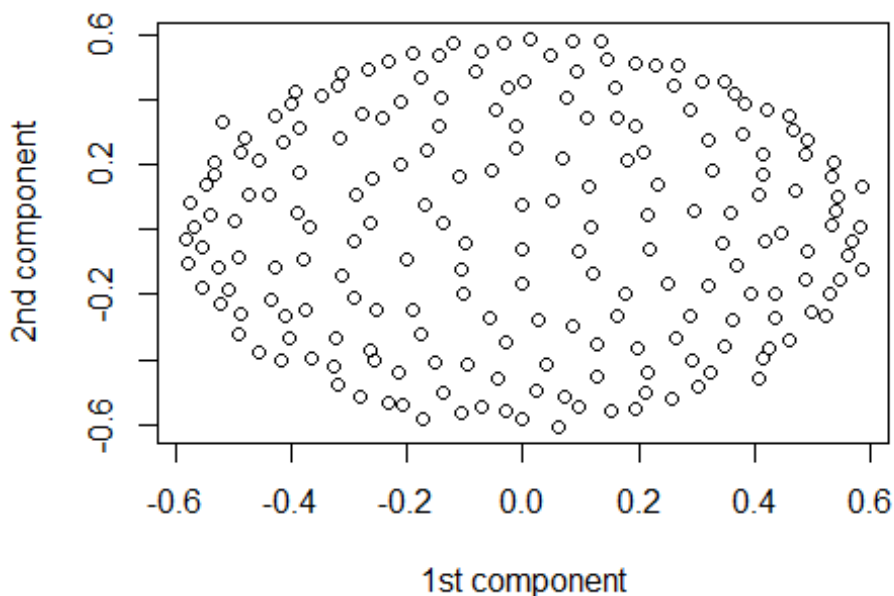
#set.seed(12345) and init <- scale(matrix(runif(m*n),ncol=m),scale=FALSE) where n represents the sample size and m represents the dimensionality of the solution. Make a plot of the two-dimensional solution. Do the results support that the data come from one homogeneous population?

```
set.seed(12345)
m <- 2
init <- scale(matrix(runif(m*n),ncol=m),scale = FALSE)
non_param_mds <- isoMDS(D,k=m,y=init)

## initial value 43.001235
## iter 5 value 41.668593
## iter 5 value 41.629957
## iter 5 value 41.629319
## final value 41.629319
## converged

plot(non_param_mds$points[,1], non_param_mds$points[,2], xlab = "1st component", y
lab = "2nd component", main="2-dimensional solution:")
```

2-dimensional solution:



In the context of applying non-parametric Multidimensional Scaling, the absence of distinct clusters in the data suggests that there are no clearly defined or discernible groups. The outcome implies that the data likely comes from an homogeneous population, with no evident patterns of separation among the data points.

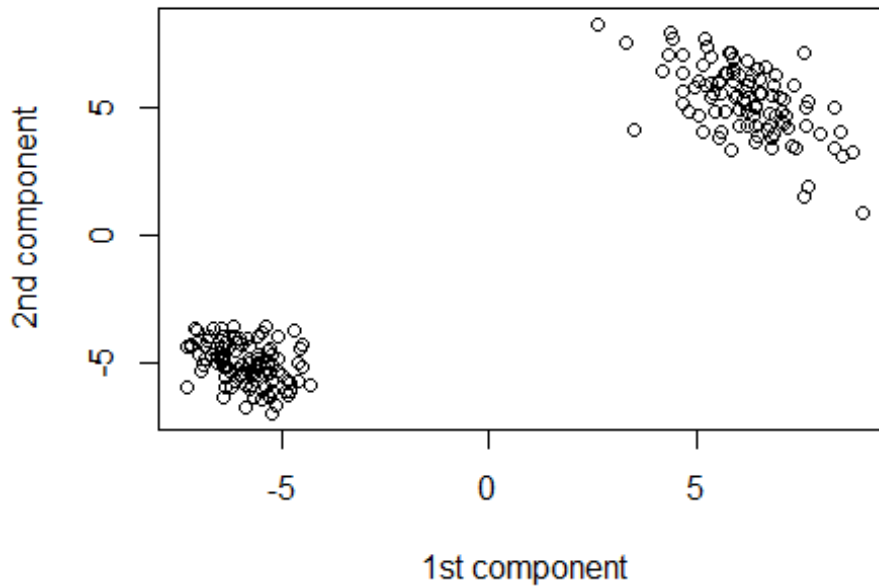
8. Try some additional runs of the two-dimensional isoMDS with different initial configurations. Make a plot of the solutions and report the STRESS for each of them. What do you observe?

```
n_iter<-10
for (i in seq(1:n_iter)){
  init <- scale(matrix(runif(m*n),ncol=m),scale=FALSE)
  non_param_mds <- isoMDS(D, y=init, k=m)
  stress <- non_param_mds$stress

  plot(non_param_mds$points[,1], non_param_mds$points[,2], xlab = "1st component",
ylab = "2nd component", main=paste("Result of Non-parametric MDS - ", i))
  cat("Stress for ", i,"th iteration: ",stress)
}

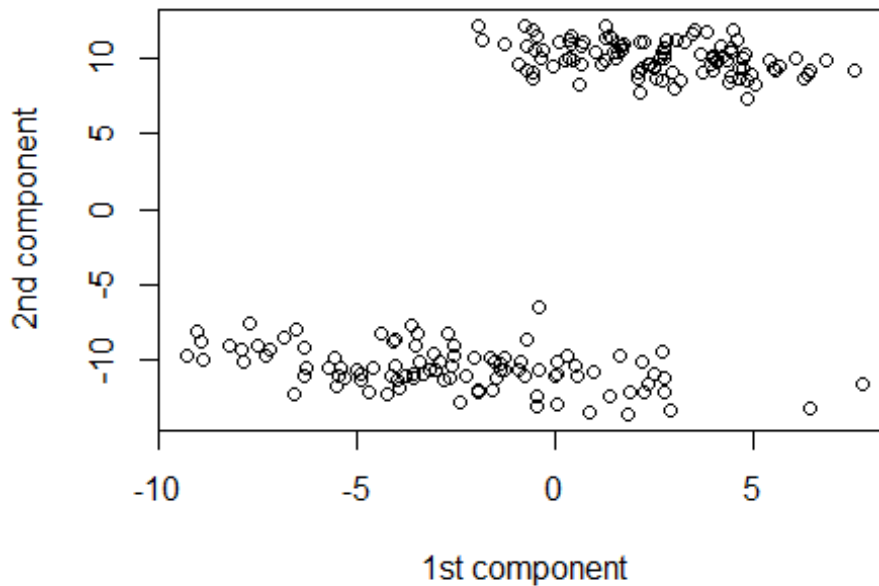
## initial   value 42.943761
## iter      5 value 41.427197
## iter     10 value 40.002996
## iter     15 value 38.060966
## iter     20 value 29.659310
## iter     25 value 21.653732
## iter     30 value 17.463856
## iter     35 value 14.418036
## iter     40 value 13.070793
## iter     45 value 12.426749
## iter     50 value 12.098786
## final    value 12.098786
## stopped after 50 iterations
```

Result of Non-parametric MDS - 1



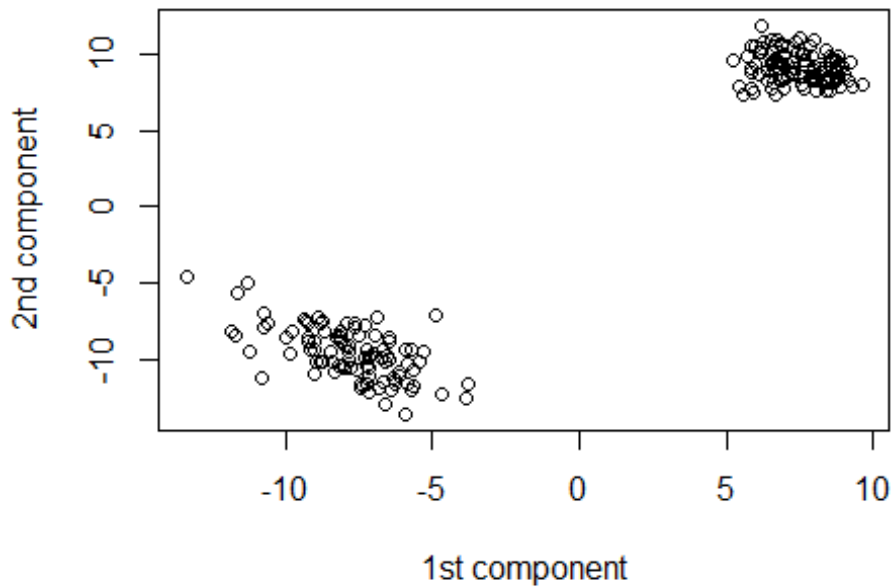
```
## Stress for 1 th iteration: 12.09879initial value 42.871770
## iter 5 value 41.643057
## iter 10 value 39.862693
## iter 15 value 39.086758
## iter 20 value 35.726119
## iter 25 value 28.329909
## iter 30 value 22.975542
## iter 35 value 19.446347
## iter 40 value 17.493380
## iter 45 value 15.388138
## iter 50 value 13.732536
## final value 13.732536
## stopped after 50 iterations
```

Result of Non-parametric MDS - 2



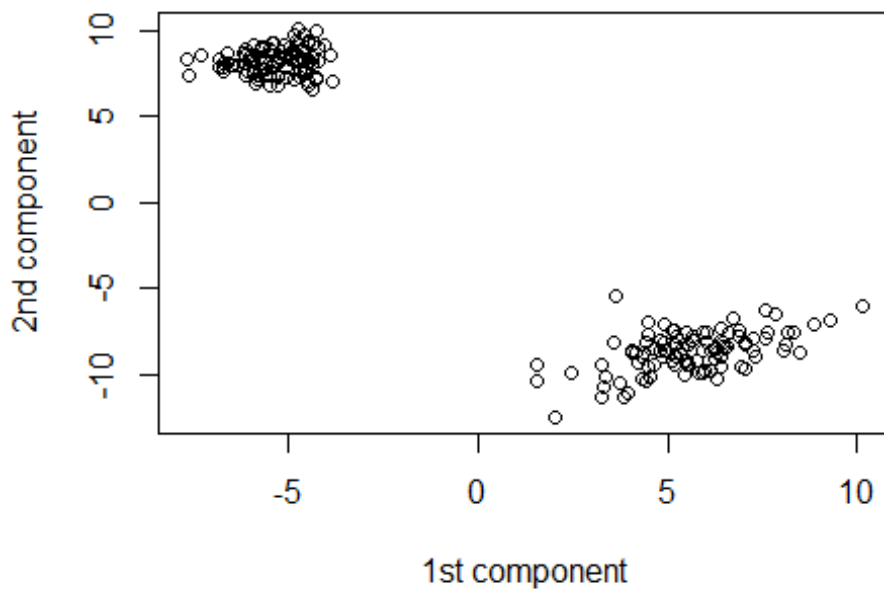
```
## Stress for 2 th iteration: 13.73254initial value 42.750591
## iter 5 value 41.580957
## iter 10 value 40.434989
## iter 15 value 38.226909
## iter 20 value 29.256158
## iter 25 value 19.959781
## iter 30 value 16.331637
## iter 35 value 13.632103
## iter 40 value 12.522541
## iter 45 value 12.111644
## final value 11.858405
## converged
```


Result of Non-parametric MDS - 3



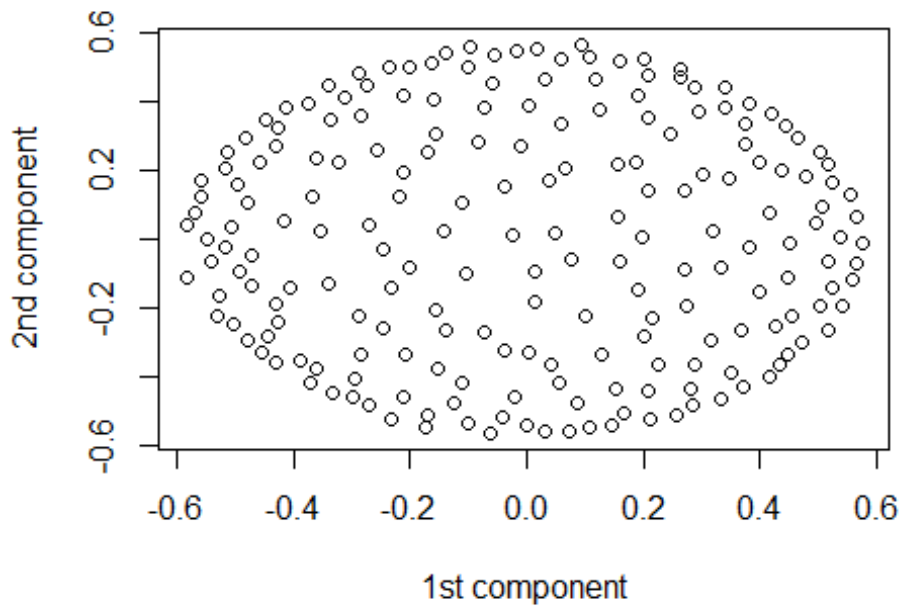
```
## Stress for 3 th iteration: 11.85841initial value 42.940739
## iter 5 value 41.589356
## iter 10 value 39.480641
## iter 15 value 35.914108
## iter 20 value 25.072993
## iter 25 value 17.130764
## iter 30 value 14.390847
## iter 35 value 12.968520
## iter 40 value 12.026409
## iter 45 value 11.666159
## final value 11.641099
## converged
```

Result of Non-parametric MDS - 4



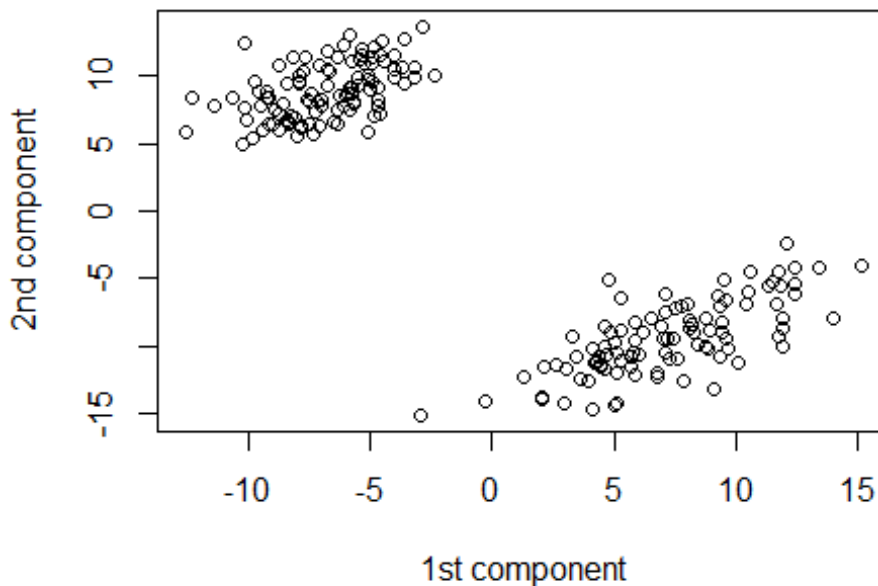
```
## Stress for 4 th iteration: 11.6411initial value 42.741937
## final value 41.686800
## converged
```

Result of Non-parametric MDS - 5



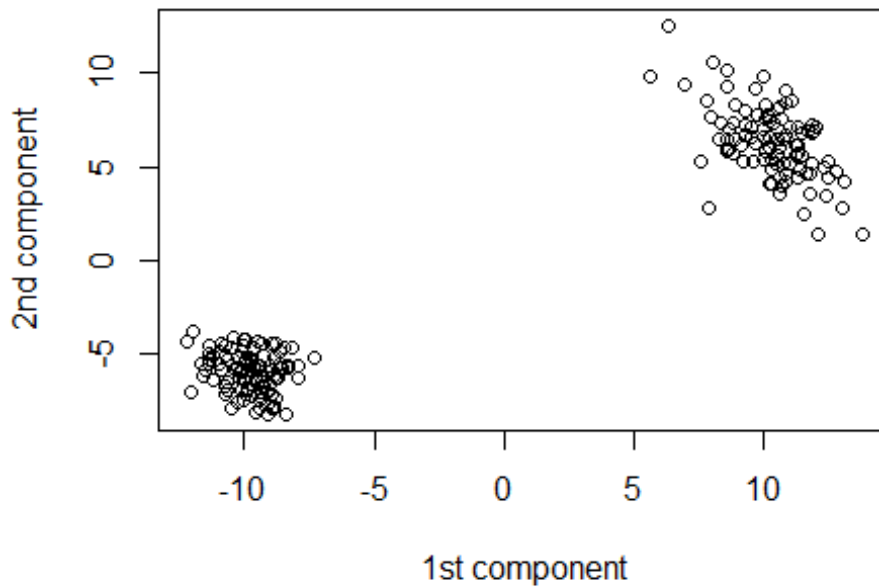
```
## Stress for 5 th iteration: 41.6868initial value 42.687586
## iter 5 value 41.291632
## iter 10 value 39.485178
## iter 15 value 39.178577
## iter 20 value 38.580936
## iter 25 value 33.047247
## iter 30 value 25.826733
## iter 35 value 21.806066
## iter 40 value 18.522455
## iter 45 value 16.421340
## iter 50 value 14.652942
## final value 14.652942
## stopped after 50 iterations
```

Result of Non-parametric MDS - 6



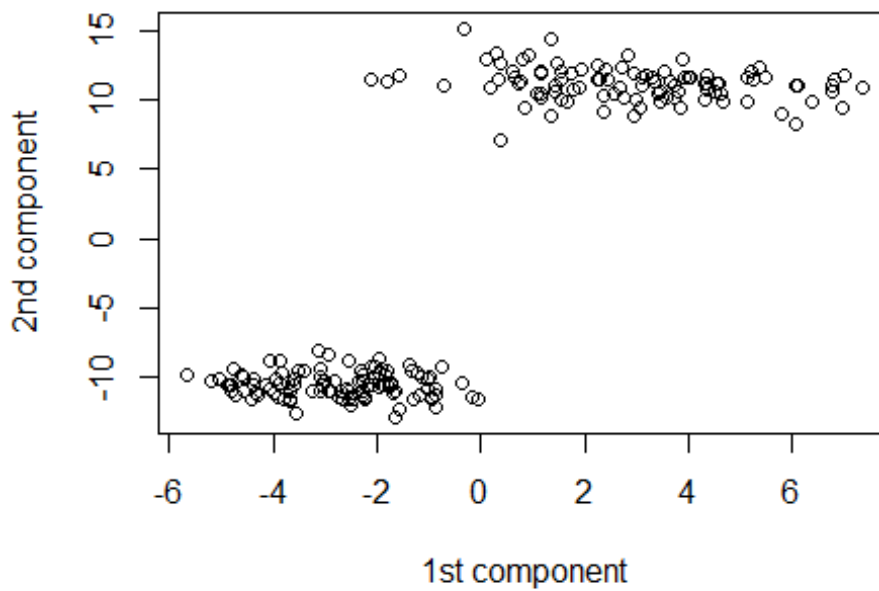
```
## Stress for 6 th iteration: 14.65294initial value 42.749593
## iter 5 value 41.590515
## iter 10 value 39.943722
## iter 15 value 38.789425
## iter 20 value 31.501041
## iter 25 value 21.182051
## iter 30 value 17.654690
## iter 35 value 13.611497
## iter 40 value 12.643228
## iter 45 value 12.129782
## iter 50 value 11.739946
## final value 11.739946
## stopped after 50 iterations
```

Result of Non-parametric MDS - 7



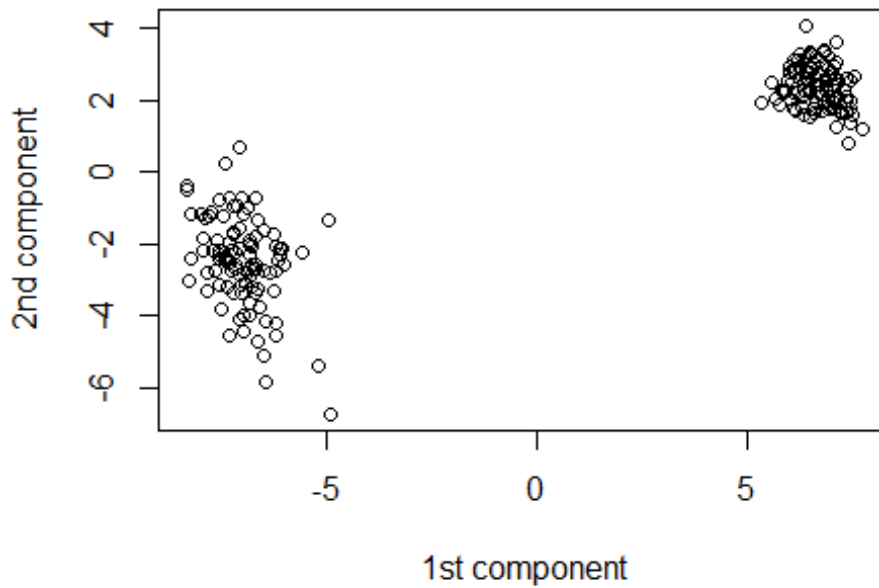
```
## Stress for 7 th iteration: 11.73995initial value 43.053665
## iter 5 value 41.485155
## iter 10 value 39.959837
## iter 15 value 39.034109
## iter 20 value 32.921849
## iter 25 value 23.583706
## iter 30 value 19.632836
## iter 35 value 16.152837
## iter 40 value 14.599163
## iter 45 value 13.076166
## iter 50 value 12.076263
## final value 12.076263
## stopped after 50 iterations
```

Result of Non-parametric MDS - 8



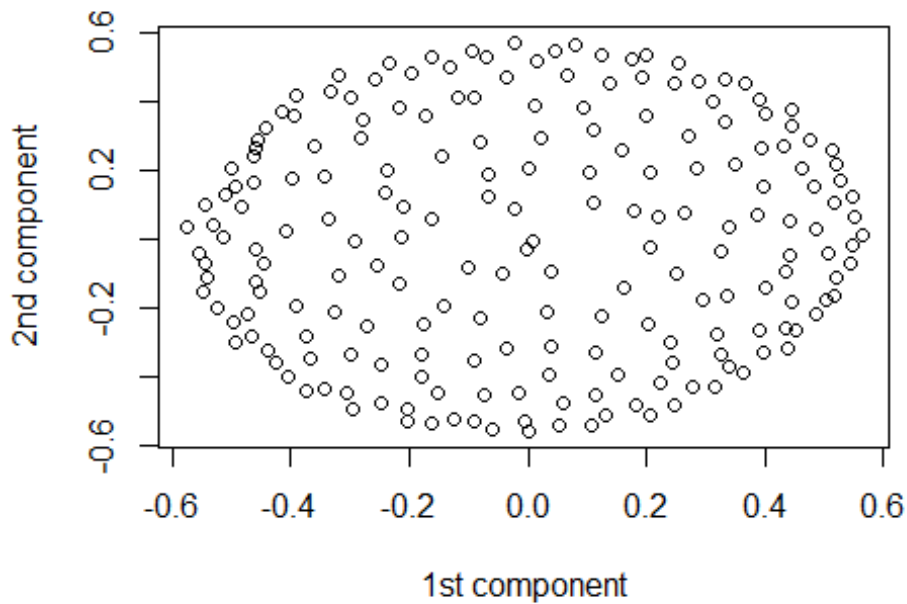
```
## Stress for 8 th iteration: 12.07626initial value 43.172234
## iter 5 value 41.629487
## iter 10 value 40.637651
## iter 15 value 28.073191
## iter 20 value 17.935615
## iter 25 value 13.762613
## iter 30 value 12.505731
## iter 35 value 12.051862
## iter 40 value 11.824039
## final value 11.724180
## converged
```

Result of Non-parametric MDS - 9



```
## Stress for 9 th iteration: 11.72418initial value 42.787973
## final value 41.714115
## converged
```

Result of Non-parametric MDS - 10



```
## Stress for 10 th iteration: 41.71411
```

What we observe is that 2 different patterns emerge in the runs using different initial configurations. Therefore, based on the initial configuration we have different outcomes if whether the population is homogeneous or not. It's worth noting that the results with 2 clusters have a lower stress.

9

9. Compute the stress for a 1, 2, 3, . . . , 50-dimensional solution. How many dimensions are necessary to obtain a good representation with a stress below 10? Make a plot of the stress against the number of dimensions.

The number of dimensions necessary to obtain a good representation is 13. Give the non-increasing behaviour of the plot, after this value the stress is getting smaller and smaller.

```
#set.seed(12345)

dimensions <- 50
stress_list <- NULL
for(i in 1:dimensions) {
  #init <- scale(matrix(runif(i*n), ncol=i), scale=FALSE)
  stress_list[i] <- isoMDS(D, k=i)$stress
}

## initial value 13.906634
## final value 13.906634
## converged
## initial value 15.185076
## final value 15.185076
## converged
## initial value 14.843856
## final value 14.843856
## converged
## initial value 14.078522
## final value 14.078522
## converged
## initial value 13.351683
## final value 13.351683
## converged
## initial value 12.525247
## final value 12.525247
## converged
## initial value 11.974448
## final value 11.974448
## converged
## initial value 11.662172
## final value 11.662172
## converged
## initial value 11.306433
## final value 11.306433
## converged
## initial value 10.617137
```

```
## final value 10.617137
## converged
## initial value 10.275785
## final value 10.275785
## converged
## initial value 10.008590
## final value 10.008590
## converged
## initial value 9.614630
## final value 9.614630
## converged
## initial value 9.108197
## final value 9.108197
## converged
## initial value 8.707142
## final value 8.707142
## converged
## initial value 8.451913
## final value 8.451913
## converged
## initial value 8.062375
## final value 8.062375
## converged
## initial value 8.047934
## final value 8.047934
## converged
## initial value 7.864988
## final value 7.864988
## converged
## initial value 7.683222
## final value 7.683222
## converged
## initial value 7.468312
## final value 7.468312
## converged
## initial value 7.218013
## final value 7.218013
## converged
## initial value 6.963368
## final value 6.963368
## converged
## initial value 6.659217
## final value 6.659217
## converged
## initial value 6.458941
## final value 6.458941
## converged
## initial value 6.233994
## final value 6.233994
## converged
## initial value 6.117082
```



```
## final value 6.117082
## converged
## initial value 5.924375
## final value 5.924375
## converged
## initial value 5.861823
## final value 5.861823
## converged
## initial value 5.610832
## final value 5.610832
## converged
## initial value 5.449734
## final value 5.449734
## converged
## initial value 5.269790
## final value 5.269790
## converged
## initial value 5.139134
## final value 5.139134
## converged
## initial value 4.996730
## final value 4.996729
## converged
## initial value 4.811151
## final value 4.811151
## converged
## initial value 4.640897
## final value 4.640897
## converged
## initial value 4.546486
## final value 4.546486
## converged
## initial value 4.390268
## final value 4.390267
## converged
## initial value 4.242612
## final value 4.242612
## converged
## initial value 4.118732
## final value 4.118732
## converged
## initial value 4.039682
## final value 4.039682
## converged
## initial value 3.939259
## final value 3.939259
## converged
## initial value 3.829421
## final value 3.829421
## converged
## initial value 3.759687
```

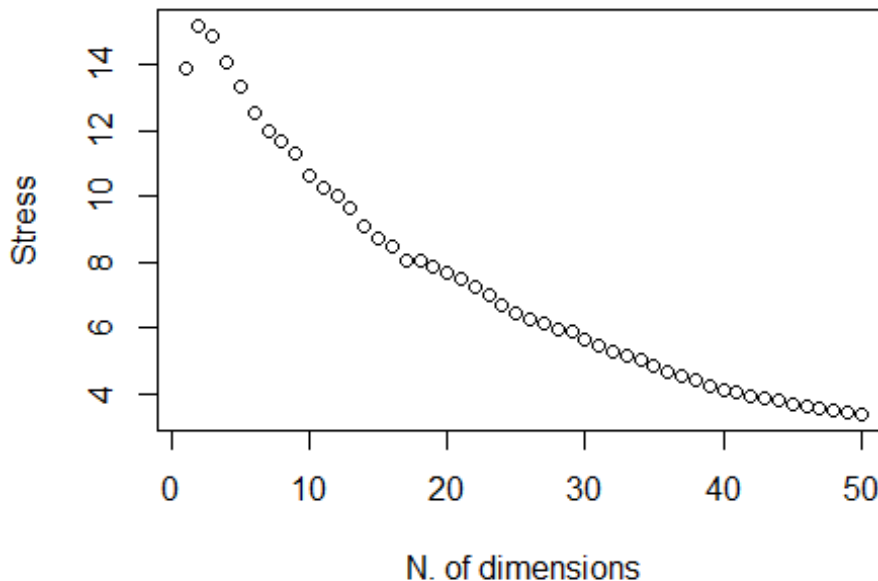
```

## final value 3.759687
## converged
## initial value 3.686426
## final value 3.686426
## converged
## initial value 3.588795
## final value 3.588795
## converged
## initial value 3.532327
## final value 3.532327
## converged
## initial value 3.471093
## final value 3.471093
## converged
## initial value 3.430112
## final value 3.430112
## converged
## initial value 3.360582
## final value 3.360582
## converged

plot(1:dimensions, stress_list, xlab="N. of dimensions", ylab="Stress", main="Stress against N. dimensions")

```

Stress against N. dimensions



```

list_dimensions<-1:50
list_dimensions[stress_list<10] #n. dimensions for stress<10

```

```
## [1] 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 31 32 33 34 35 36 37
## [26] 38 39 40 41 42 43 44 45 46 47 48 49 50
```

10

10. Run the two-dimensional isoMDS a hundred times, each time using a different random initial configuration using the instructions above. Report the stress of the best and the worse run, and plot the corresponding maps. Compare your results to the metric MDS and comment on your findings.

Stress of worst run: 42.90679 Stress of best run: 11.43862

```
n_iter <- 100
stress_list <- NULL
non_params_mds_list <- list()

for (i in seq(1:n_iter)) {
  init <- scale(matrix(runif(m * n), ncol = m), scale = FALSE)
  non_param_mds <- isoMDS(D, y = init, k = m)
  stress_list[i] <- non_param_mds$stress
  non_params_mds_list[[i]] <- non_param_mds

  cat("Stress for ", i, "th iteration: ", stress_list[i], "\n")
}

## initial value 43.564259
## iter 5 value 41.543462
## iter 10 value 40.032698
## iter 15 value 37.982675
## iter 20 value 30.530080
## iter 25 value 22.857289
## iter 30 value 18.045406
## iter 35 value 14.378411
## iter 40 value 13.640402
## iter 45 value 12.869677
## iter 50 value 11.984643
## final value 11.984643
## stopped after 50 iterations
## Stress for 1 th iteration: 11.98464
## initial value 43.154912
## iter 5 value 41.656934
## iter 5 value 41.626314
## iter 5 value 41.623995
## final value 41.623995
## converged
## Stress for 2 th iteration: 41.62399
## initial value 42.847441
## final value 41.647872
## converged
## Stress for 3 th iteration: 41.64787
## initial value 42.441838
```

```
## final value 41.668189
## converged
## Stress for 4 th iteration: 41.66819
## initial value 42.547112
## iter 5 value 41.544287
## iter 10 value 40.156972
## iter 15 value 37.103475
## iter 20 value 25.608132
## iter 25 value 19.171616
## iter 30 value 14.716695
## iter 35 value 12.958316
## iter 40 value 11.994017
## iter 45 value 11.656369
## iter 45 value 11.649638
## iter 45 value 11.645642
## final value 11.645642
## converged
## Stress for 5 th iteration: 11.64564
## initial value 43.488921
## iter 5 value 41.573662
## iter 10 value 39.802111
## iter 15 value 38.342548
## iter 20 value 29.262644
## iter 25 value 19.134296
## iter 30 value 16.227120
## iter 35 value 13.250100
## iter 40 value 12.310201
## iter 45 value 12.071257
## final value 11.734351
## converged
## Stress for 6 th iteration: 11.73435
## initial value 42.694560
## iter 5 value 41.679947
## iter 5 value 41.674565
## iter 5 value 41.674565
## final value 41.674565
## converged
## Stress for 7 th iteration: 41.67456
## initial value 43.234830
## final value 41.663354
## converged
## Stress for 8 th iteration: 41.66335
## initial value 42.738511
## iter 5 value 40.729410
## iter 10 value 39.503204
## iter 15 value 37.027112
## iter 20 value 30.883757
## iter 25 value 25.296826
## iter 30 value 20.882295
## iter 35 value 17.378017
## iter 40 value 16.469532
```

```
## iter 45 value 14.126420
## iter 50 value 12.548321
## final value 12.548321
## stopped after 50 iterations
## Stress for 9 th iteration: 12.54832
## initial value 42.868715
## iter 5 value 41.518366
## iter 10 value 39.297759
## iter 15 value 31.073124
## iter 20 value 20.430374
## iter 25 value 13.948410
## iter 30 value 12.403626
## iter 35 value 11.893926
## final value 11.671211
## converged
## Stress for 10 th iteration: 11.67121
## initial value 43.341356
## final value 42.197447
## converged
## Stress for 11 th iteration: 42.19745
## initial value 42.697035
## final value 41.667837
## converged
## Stress for 12 th iteration: 41.66784
## initial value 42.794806
## iter 5 value 41.670063
## final value 41.564516
## converged
## Stress for 13 th iteration: 41.56452
## initial value 42.818422
## iter 5 value 41.691132
## iter 5 value 41.670627
## iter 5 value 41.670627
## final value 41.670627
## converged
## Stress for 14 th iteration: 41.67063
## initial value 42.512275
## iter 5 value 41.401087
## iter 10 value 39.657915
## iter 15 value 37.366792
## iter 20 value 30.578684
## iter 25 value 25.497959
## iter 30 value 19.782798
## iter 35 value 15.544443
## iter 40 value 14.668594
## iter 45 value 13.330450
## iter 50 value 12.714543
## final value 12.714543
## stopped after 50 iterations
## Stress for 15 th iteration: 12.71454
## initial value 42.617771
```

```
## iter    5 value 41.592516
## final   value 41.477708
## converged
## Stress for 16 th iteration: 41.47771
## initial value 42.883990
## final   value 41.684567
## converged
## Stress for 17 th iteration: 41.68457
## initial value 42.849136
## iter    5 value 41.265947
## iter   10 value 39.684725
## iter   15 value 38.708275
## iter   20 value 31.844894
## iter   25 value 22.903387
## iter   30 value 18.998567
## iter   35 value 14.852591
## iter   40 value 13.415760
## iter   45 value 12.573658
## iter   50 value 12.052875
## final   value 12.052875
## stopped after 50 iterations
## Stress for 18 th iteration: 12.05287
## initial value 42.586987
## final   value 41.674662
## converged
## Stress for 19 th iteration: 41.67466
## initial value 42.458387
## final   value 41.692680
## converged
## Stress for 20 th iteration: 41.69268
## initial value 42.613554
## final   value 41.645665
## converged
## Stress for 21 th iteration: 41.64567
## initial value 42.901908
## iter    5 value 41.578556
## iter   10 value 39.757976
## iter   15 value 38.602719
## iter   20 value 33.328801
## iter   25 value 26.778851
## iter   30 value 21.754625
## iter   35 value 16.811188
## iter   40 value 15.419023
## iter   45 value 13.385526
## iter   50 value 12.611282
## final   value 12.611282
## stopped after 50 iterations
## Stress for 22 th iteration: 12.61128
## initial value 42.839946
## final   value 41.684079
## converged
```

```
## Stress for 23 th iteration: 41.68408
## initial value 42.831669
## iter 5 value 41.522515
## iter 10 value 39.728382
## iter 15 value 36.909406
## iter 20 value 26.785712
## iter 25 value 18.732870
## iter 30 value 14.805752
## iter 35 value 13.131675
## iter 40 value 12.058027
## final value 11.753963
## converged
## Stress for 24 th iteration: 11.75396
## initial value 43.558389
## iter 5 value 41.631028
## iter 10 value 39.946889
## iter 15 value 37.271873
## iter 20 value 27.459453
## iter 25 value 17.728723
## iter 30 value 14.477758
## iter 35 value 12.409980
## iter 40 value 11.786684
## iter 45 value 11.654818
## final value 11.630305
## converged
## Stress for 25 th iteration: 11.63031
## initial value 42.704618
## final value 41.672501
## converged
## Stress for 26 th iteration: 41.6725
## initial value 42.952311
## final value 42.949535
## converged
## Stress for 27 th iteration: 42.94953
## initial value 42.741360
## iter 5 value 41.207473
## iter 10 value 39.485467
## iter 15 value 38.993369
## iter 20 value 33.128542
## iter 25 value 24.117494
## iter 30 value 19.541595
## iter 35 value 16.921086
## iter 40 value 15.593074
## iter 45 value 14.796532
## iter 50 value 13.955516
## final value 13.955516
## stopped after 50 iterations
## Stress for 28 th iteration: 13.95552
## initial value 43.117495
## iter 5 value 41.664809
## iter 5 value 41.633311
```

```
## iter    5 value 41.632936
## final   value 41.632936
## converged
## Stress for 29 th iteration: 41.63294
## initial value 43.017313
## final   value 41.678412
## converged
## Stress for 30 th iteration: 41.67841
## initial value 42.532853
## iter    5 value 41.672848
## iter    5 value 41.653039
## iter    5 value 41.652831
## final   value 41.652831
## converged
## Stress for 31 th iteration: 41.65283
## initial value 42.917490
## iter    5 value 41.622532
## iter   10 value 39.895693
## iter   15 value 38.377814
## iter   20 value 25.506854
## iter   25 value 18.677537
## iter   30 value 15.931754
## iter   35 value 14.160077
## iter   40 value 12.609629
## iter   45 value 12.263492
## iter   50 value 11.854827
## final   value 11.854827
## stopped after 50 iterations
## Stress for 32 th iteration: 11.85483
## initial value 43.370288
## final   value 42.075400
## converged
## Stress for 33 th iteration: 42.0754
## initial value 42.851040
## iter    5 value 41.429750
## iter   10 value 39.548679
## iter   15 value 38.414506
## iter   20 value 27.225152
## iter   25 value 18.284947
## iter   30 value 13.648598
## iter   35 value 12.746153
## iter   40 value 11.812511
## iter   45 value 11.571602
## final   value 11.538144
## converged
## Stress for 34 th iteration: 11.53814
## initial value 42.874068
## iter    5 value 41.703140
## iter    5 value 41.670168
## iter    5 value 41.670168
## final   value 41.670168
```



```
## converged
## Stress for 35 th iteration: 41.67017
## initial value 42.907898
## iter 5 value 41.447682
## iter 10 value 39.585102
## iter 15 value 27.391815
## iter 20 value 17.017077
## iter 25 value 13.512567
## iter 30 value 12.327437
## iter 35 value 11.792758
## iter 40 value 11.672541
## final value 11.646659
## converged
## Stress for 36 th iteration: 11.64666
## initial value 43.053845
## iter 5 value 41.591973
## iter 10 value 39.648945
## iter 15 value 39.343855
## iter 20 value 37.402563
## iter 25 value 30.088321
## iter 30 value 22.908649
## iter 35 value 19.949935
## iter 40 value 17.523179
## iter 45 value 15.824363
## iter 50 value 13.712416
## final value 13.712416
## stopped after 50 iterations
## Stress for 37 th iteration: 13.71242
## initial value 42.751412
## iter 5 value 41.477124
## iter 10 value 39.660142
## iter 15 value 39.232727
## iter 20 value 37.987884
## iter 25 value 27.979224
## iter 30 value 20.834097
## iter 35 value 17.029963
## iter 40 value 15.885244
## iter 45 value 15.084569
## iter 50 value 13.493780
## final value 13.493780
## stopped after 50 iterations
## Stress for 38 th iteration: 13.49378
## initial value 42.722107
## iter 5 value 41.399542
## iter 10 value 37.514367
## iter 15 value 22.034550
## iter 20 value 14.773718
## iter 25 value 12.385977
## iter 30 value 11.726294
## iter 35 value 11.592924
## final value 11.559267
```

```
## converged
## Stress for 39 th iteration: 11.55927
## initial value 42.599590
## iter 5 value 41.365486
## iter 10 value 39.543764
## iter 15 value 38.581975
## iter 20 value 31.755754
## iter 25 value 23.222836
## iter 30 value 18.393975
## iter 35 value 16.496615
## iter 40 value 15.642166
## iter 45 value 14.751913
## iter 50 value 13.246714
## final value 13.246714
## stopped after 50 iterations
## Stress for 40 th iteration: 13.24671
## initial value 42.976891
## iter 5 value 41.582836
## iter 10 value 39.908388
## iter 15 value 39.182371
## iter 20 value 34.144425
## iter 25 value 24.150126
## iter 30 value 18.581022
## iter 35 value 15.872097
## iter 40 value 13.749784
## iter 45 value 12.592987
## iter 50 value 12.123582
## final value 12.123582
## stopped after 50 iterations
## Stress for 41 th iteration: 12.12358
## initial value 43.199547
## final value 42.972384
## converged
## Stress for 42 th iteration: 42.97238
## initial value 43.070101
## iter 5 value 41.292975
## iter 10 value 39.463295
## iter 15 value 36.406763
## iter 20 value 24.589776
## iter 25 value 17.094436
## iter 30 value 13.871259
## iter 35 value 12.820029
## iter 40 value 12.247055
## iter 45 value 11.974467
## iter 50 value 11.712635
## final value 11.712635
## stopped after 50 iterations
## Stress for 43 th iteration: 11.71264
## initial value 42.767876
## final value 42.646690
## converged
```

```
## Stress for 44 th iteration: 42.64669
## initial value 43.115372
## iter 5 value 41.511602
## iter 10 value 39.682913
## iter 15 value 38.497668
## iter 20 value 30.000041
## iter 25 value 22.254210
## iter 30 value 18.784033
## iter 35 value 15.081843
## iter 40 value 13.002703
## iter 45 value 12.422926
## iter 50 value 11.924657
## final value 11.924657
## stopped after 50 iterations
## Stress for 45 th iteration: 11.92466
## initial value 42.668020
## iter 5 value 41.472070
## iter 10 value 39.505689
## iter 15 value 34.048871
## iter 20 value 24.405753
## iter 25 value 18.382596
## iter 30 value 15.067207
## iter 35 value 13.426011
## iter 40 value 12.125235
## iter 45 value 11.938224
## final value 11.668064
## converged
## Stress for 46 th iteration: 11.66806
## initial value 43.095457
## final value 42.204197
## converged
## Stress for 47 th iteration: 42.2042
## initial value 43.442766
## iter 5 value 41.695998
## iter 5 value 41.658179
## iter 5 value 41.658179
## final value 41.658179
## converged
## Stress for 48 th iteration: 41.65818
## initial value 43.067472
## iter 5 value 41.679600
## iter 5 value 41.665518
## iter 5 value 41.665518
## final value 41.665518
## converged
## Stress for 49 th iteration: 41.66552
## initial value 42.764107
## iter 5 value 41.418358
## iter 10 value 39.509786
## iter 15 value 30.743161
## iter 20 value 19.320663
```

```
## iter 25 value 13.896249
## iter 30 value 12.420624
## iter 35 value 11.901089
## final value 11.687951
## converged
## Stress for 50 th iteration: 11.68795
## initial value 42.491769
## iter 5 value 41.574626
## iter 10 value 40.778843
## iter 15 value 34.249919
## iter 20 value 23.559608
## iter 25 value 16.256645
## iter 30 value 14.177083
## iter 35 value 12.459387
## iter 40 value 12.164171
## iter 45 value 11.816949
## iter 45 value 11.807874
## iter 45 value 11.801831
## final value 11.801831
## converged
## Stress for 51 th iteration: 11.80183
## initial value 42.617879
## final value 41.683884
## converged
## Stress for 52 th iteration: 41.68388
## initial value 42.910124
## iter 5 value 41.653123
## iter 10 value 40.914907
## iter 15 value 32.840274
## iter 20 value 21.587321
## iter 25 value 15.285608
## iter 30 value 12.996897
## iter 35 value 12.200033
## iter 40 value 11.903554
## final value 11.704327
## converged
## Stress for 53 th iteration: 11.70433
## initial value 43.073653
## final value 42.020934
## converged
## Stress for 54 th iteration: 42.02093
## initial value 42.511143
## iter 5 value 41.588905
## iter 10 value 40.826899
## iter 15 value 34.861055
## iter 20 value 24.858867
## iter 25 value 16.895954
## iter 30 value 13.194124
## iter 35 value 12.061268
## iter 40 value 11.680442
## final value 11.637569
```

```
## converged
## Stress for 55 th iteration: 11.63757
## initial value 43.131124
## iter 5 value 41.629492
## iter 10 value 40.883322
## iter 15 value 39.074746
## iter 20 value 30.700506
## iter 25 value 19.876139
## iter 30 value 15.904782
## iter 35 value 13.922370
## iter 40 value 12.888413
## iter 45 value 12.347407
## iter 50 value 12.153927
## final value 12.153927
## stopped after 50 iterations
## Stress for 56 th iteration: 12.15393
## initial value 42.766132
## iter 5 value 41.627248
## iter 10 value 39.770663
## iter 15 value 37.927155
## iter 20 value 29.248022
## iter 25 value 21.577430
## iter 30 value 17.775906
## iter 35 value 14.518345
## iter 40 value 13.219674
## iter 45 value 12.531649
## final value 11.767400
## converged
## Stress for 57 th iteration: 11.7674
## initial value 42.505458
## final value 41.678323
## converged
## Stress for 58 th iteration: 41.67832
## initial value 43.050272
## iter 5 value 41.683914
## iter 5 value 41.658027
## iter 5 value 41.657956
## final value 41.657956
## converged
## Stress for 59 th iteration: 41.65796
## initial value 42.983531
## iter 5 value 41.541331
## iter 10 value 41.118910
## iter 15 value 39.290779
## iter 20 value 31.755319
## iter 25 value 21.784170
## iter 30 value 17.868773
## iter 35 value 14.555643
## iter 40 value 13.372594
## iter 45 value 12.441419
## iter 50 value 11.871535
```

```
## final value 11.871535
## stopped after 50 iterations
## Stress for 60 th iteration: 11.87154
## initial value 42.888072
## iter 5 value 41.643594
## iter 10 value 41.143197
## iter 15 value 39.017200
## iter 20 value 30.746888
## iter 25 value 20.413256
## iter 30 value 15.241800
## iter 35 value 13.058047
## iter 40 value 12.349343
## iter 45 value 11.829650
## iter 50 value 11.645683
## final value 11.645683
## stopped after 50 iterations
## Stress for 61 th iteration: 11.64568
## initial value 43.473279
## iter 5 value 41.562818
## iter 10 value 40.321377
## iter 15 value 37.752185
## iter 20 value 25.187587
## iter 25 value 16.462965
## iter 30 value 12.971973
## iter 35 value 12.162608
## iter 40 value 11.698631
## iter 45 value 11.548895
## final value 11.483701
## converged
## Stress for 62 th iteration: 11.4837
## initial value 42.955518
## iter 5 value 41.674600
## iter 5 value 41.634536
## iter 5 value 41.629677
## final value 41.629677
## converged
## Stress for 63 th iteration: 41.62968
## initial value 42.505686
## final value 41.672933
## converged
## Stress for 64 th iteration: 41.67293
## initial value 43.342419
## iter 5 value 41.684078
## iter 5 value 41.676996
## iter 5 value 41.676996
## final value 41.676996
## converged
## Stress for 65 th iteration: 41.677
## initial value 43.123031
## iter 5 value 41.675655
## iter 5 value 41.650914
```

```
## iter    5 value 41.650642
## final   value 41.650642
## converged
## Stress for 66 th iteration: 41.65064
## initial value 42.742927
## iter    5 value 41.463794
## iter   10 value 38.440389
## iter   15 value 23.195298
## iter   20 value 13.436598
## iter   25 value 11.938178
## iter   30 value 11.666857
## iter   35 value 11.556916
## final   value 11.494065
## converged
## Stress for 67 th iteration: 11.49407
## initial value 42.668951
## iter    5 value 41.657405
## iter   10 value 40.732403
## iter   15 value 39.285083
## iter   20 value 36.393152
## iter   25 value 29.904427
## iter   30 value 20.957130
## iter   35 value 16.962014
## iter   40 value 14.408381
## iter   45 value 13.906032
## iter   50 value 12.202272
## final   value 12.202272
## stopped after 50 iterations
## Stress for 68 th iteration: 12.20227
## initial value 42.790421
## iter    5 value 41.549618
## iter   10 value 39.806957
## iter   15 value 39.175412
## iter   20 value 35.597693
## iter   25 value 26.877423
## iter   30 value 20.815572
## iter   35 value 17.631127
## iter   40 value 16.271181
## iter   45 value 15.010125
## iter   50 value 13.403397
## final   value 13.403397
## stopped after 50 iterations
## Stress for 69 th iteration: 13.4034
## initial value 42.813294
## iter    5 value 41.664897
## iter    5 value 41.630099
## iter    5 value 41.628571
## final   value 41.628571
## converged
## Stress for 70 th iteration: 41.62857
## initial value 42.638796
```

```
## iter    5 value 41.456723
## iter   10 value 39.426700
## iter   15 value 32.352647
## iter   20 value 22.052027
## iter   25 value 15.131286
## iter   30 value 12.225224
## iter   35 value 11.798223
## iter   40 value 11.667291
## final  value 11.633369
## converged
## Stress for 71 th iteration: 11.63337
## initial value 42.850635
## iter    5 value 41.643887
## iter   10 value 41.348925
## iter   15 value 39.538339
## iter   20 value 32.169509
## iter   25 value 19.497408
## iter   30 value 14.415617
## iter   35 value 12.851811
## iter   40 value 12.168866
## iter   45 value 11.896734
## final  value 11.738899
## converged
## Stress for 72 th iteration: 11.7389
## initial value 43.516729
## iter    5 value 41.647869
## iter    5 value 41.623119
## iter    5 value 41.623119
## final  value 41.623119
## converged
## Stress for 73 th iteration: 41.62312
## initial value 43.089146
## final  value 41.706210
## converged
## Stress for 74 th iteration: 41.70621
## initial value 42.530999
## iter    5 value 41.686357
## iter    5 value 41.671314
## iter    5 value 41.671314
## final  value 41.671314
## converged
## Stress for 75 th iteration: 41.67131
## initial value 42.786220
## iter    5 value 41.398146
## iter    5 value 41.365893
## iter   10 value 39.745445
## iter   15 value 39.328885
## iter   20 value 38.704478
## iter   25 value 27.424491
## iter   30 value 17.625610
## iter   35 value 13.844263
```



```
## iter 40 value 12.473521
## iter 45 value 11.749001
## iter 50 value 11.611477
## final value 11.611477
## stopped after 50 iterations
## Stress for 76 th iteration: 11.61148
## initial value 42.751946
## iter 5 value 41.518733
## iter 10 value 39.559496
## iter 15 value 35.835554
## iter 20 value 25.869862
## iter 25 value 17.666688
## iter 30 value 13.976113
## iter 35 value 12.836939
## iter 40 value 12.234038
## iter 45 value 11.932432
## final value 11.734118
## converged
## Stress for 77 th iteration: 11.73412
## initial value 43.242020
## iter 5 value 41.688557
## iter 5 value 41.673446
## iter 5 value 41.673446
## final value 41.673446
## converged
## Stress for 78 th iteration: 41.67345
## initial value 42.970185
## iter 5 value 41.316492
## iter 10 value 39.181119
## iter 15 value 32.772307
## iter 20 value 24.137440
## iter 25 value 16.238509
## iter 30 value 13.769250
## iter 35 value 12.388229
## iter 40 value 11.999669
## final value 11.676714
## converged
## Stress for 79 th iteration: 11.67671
## initial value 42.830224
## iter 5 value 40.935422
## iter 10 value 39.138860
## iter 15 value 34.372449
## iter 20 value 25.512514
## iter 25 value 16.303621
## iter 30 value 14.706918
## iter 35 value 12.507738
## iter 40 value 12.087496
## iter 45 value 11.905595
## final value 11.717343
## converged
## Stress for 80 th iteration: 11.71734
```

```
## initial value 42.779989
## iter 5 value 41.542890
## iter 10 value 39.490864
## iter 15 value 37.082882
## iter 20 value 30.202408
## iter 25 value 24.167591
## iter 30 value 20.447899
## iter 35 value 17.211609
## iter 40 value 14.447102
## iter 45 value 13.093839
## iter 50 value 11.876137
## final value 11.876137
## stopped after 50 iterations
## Stress for 81 th iteration: 11.87614
## initial value 42.741804
## iter 5 value 41.689965
## iter 5 value 41.681337
## iter 5 value 41.681337
## final value 41.681337
## converged
## Stress for 82 th iteration: 41.68134
## initial value 42.764531
## final value 41.641021
## converged
## Stress for 83 th iteration: 41.64102
## initial value 42.845060
## iter 5 value 41.501059
## iter 10 value 40.064799
## iter 15 value 39.176944
## iter 20 value 36.708657
## iter 25 value 28.917963
## iter 30 value 23.065923
## iter 35 value 18.572192
## iter 40 value 16.813176
## iter 45 value 14.433141
## iter 50 value 13.348179
## final value 13.348179
## stopped after 50 iterations
## Stress for 84 th iteration: 13.34818
## initial value 42.974752
## iter 5 value 41.553990
## iter 10 value 40.319523
## iter 15 value 39.150809
## iter 20 value 35.315126
## iter 25 value 27.431629
## iter 30 value 20.784361
## iter 35 value 16.309524
## iter 40 value 14.966538
## iter 45 value 13.735357
## iter 50 value 12.524443
## final value 12.524443
```

```
## stopped after 50 iterations
## Stress for 85 th iteration: 12.52444
## initial value 42.911476
## final value 41.693425
## converged
## Stress for 86 th iteration: 41.69342
## initial value 42.555177
## iter 5 value 41.137447
## iter 10 value 39.426768
## iter 15 value 36.008708
## iter 20 value 28.417994
## iter 25 value 20.889961
## iter 30 value 17.880288
## iter 35 value 14.423375
## iter 40 value 13.668810
## iter 45 value 12.485955
## iter 50 value 11.820043
## final value 11.820043
## stopped after 50 iterations
## Stress for 87 th iteration: 11.82004
## initial value 42.988933
## final value 41.667139
## converged
## Stress for 88 th iteration: 41.66714
## initial value 42.622388
## iter 5 value 41.602089
## iter 10 value 40.095073
## iter 15 value 38.532978
## iter 20 value 29.127832
## iter 25 value 20.706735
## iter 30 value 17.341126
## iter 35 value 14.242974
## iter 40 value 13.146108
## iter 45 value 12.421764
## iter 50 value 11.943074
## final value 11.943074
## stopped after 50 iterations
## Stress for 89 th iteration: 11.94307
## initial value 43.062676
## iter 5 value 41.648695
## iter 10 value 41.315186
## iter 15 value 34.236052
## iter 20 value 17.881158
## iter 25 value 13.434041
## iter 30 value 12.222659
## iter 35 value 11.849403
## iter 40 value 11.736610
## final value 11.679134
## converged
## Stress for 90 th iteration: 11.67913
## initial value 42.828198
```

```
## iter    5 value 41.059144
## iter   10 value 39.270591
## iter   15 value 34.926100
## iter   20 value 27.984556
## iter   25 value 22.051742
## iter   30 value 16.977004
## iter   35 value 14.805640
## iter   40 value 13.289459
## iter   45 value 12.120086
## iter   50 value 11.778131
## final   value 11.778131
## stopped after 50 iterations
## Stress for 91 th iteration: 11.77813
## initial value 42.549231
## final   value 42.031624
## converged
## Stress for 92 th iteration: 42.03162
## initial value 43.022650
## iter    5 value 41.664095
## iter    5 value 41.623708
## iter    5 value 41.623188
## final   value 41.623188
## converged
## Stress for 93 th iteration: 41.62319
## initial value 42.750377
## iter    5 value 41.635996
## iter   10 value 41.026187
## iter   15 value 38.988838
## iter   20 value 30.758468
## iter   25 value 22.513293
## iter   30 value 16.530867
## iter   35 value 14.122990
## iter   40 value 12.981354
## iter   45 value 12.458543
## iter   50 value 12.010741
## final   value 12.010741
## stopped after 50 iterations
## Stress for 94 th iteration: 12.01074
## initial value 42.430652
## iter    5 value 41.227456
## iter   10 value 39.132013
## iter   15 value 29.283460
## iter   20 value 20.333394
## iter   25 value 15.352267
## iter   30 value 13.076396
## iter   35 value 12.096127
## iter   40 value 11.674610
## iter   45 value 11.565337
## final   value 11.526704
## converged
## Stress for 95 th iteration: 11.5267
```

```
## initial value 42.936950
## iter 5 value 41.655262
## iter 5 value 41.638172
## iter 5 value 41.638099
## final value 41.638099
## converged
## Stress for 96 th iteration: 41.6381
## initial value 42.728200
## iter 5 value 41.572028
## iter 10 value 39.639192
## iter 15 value 37.558899
## iter 20 value 24.314829
## iter 25 value 16.342264
## iter 30 value 13.597598
## iter 35 value 12.232831
## iter 40 value 11.880618
## final value 11.651820
## converged
## Stress for 97 th iteration: 11.65182
## initial value 43.114456
## iter 5 value 41.515694
## iter 10 value 39.790691
## iter 15 value 37.860198
## iter 20 value 28.663161
## iter 25 value 20.146571
## iter 30 value 15.464908
## iter 35 value 12.005909
## iter 40 value 11.688739
## final value 11.581079
## converged
## Stress for 98 th iteration: 11.58108
## initial value 42.698949
## iter 5 value 41.626810
## iter 10 value 40.920406
## iter 15 value 39.551196
## iter 20 value 38.508583
## iter 25 value 25.706182
## iter 30 value 18.515358
## iter 35 value 15.430208
## iter 40 value 13.414929
## iter 45 value 12.628183
## iter 50 value 12.130910
## final value 12.130910
## stopped after 50 iterations
## Stress for 99 th iteration: 12.13091
## initial value 42.741704
## final value 41.680981
## converged
## Stress for 100 th iteration: 41.68098
```

```

list_runs <- 1:n_iter
worst_run <- list_runs[stress_list == max(stress_list)]
best_run <- list_runs[stress_list == min(stress_list)]

cat("Stress of worst run: ", stress_list[worst_run])

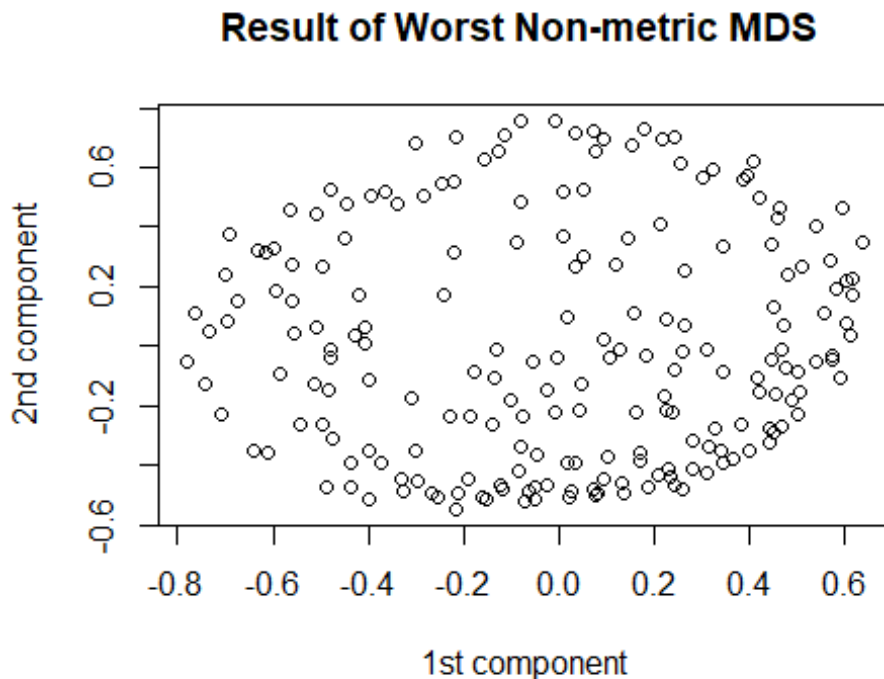
## Stress of worst run: 42.97238

cat("Stress of best run: ", stress_list[best_run])

## Stress of best run: 11.4837

# plot worst run representation
plot(non_params_mds_list[[worst_run]]$points[, 1], non_params_mds_list[[worst_run]]$points[, 2],
      xlab = "1st component", ylab = "2nd component", main = "Result of Worst Non-metric MDS ")

```

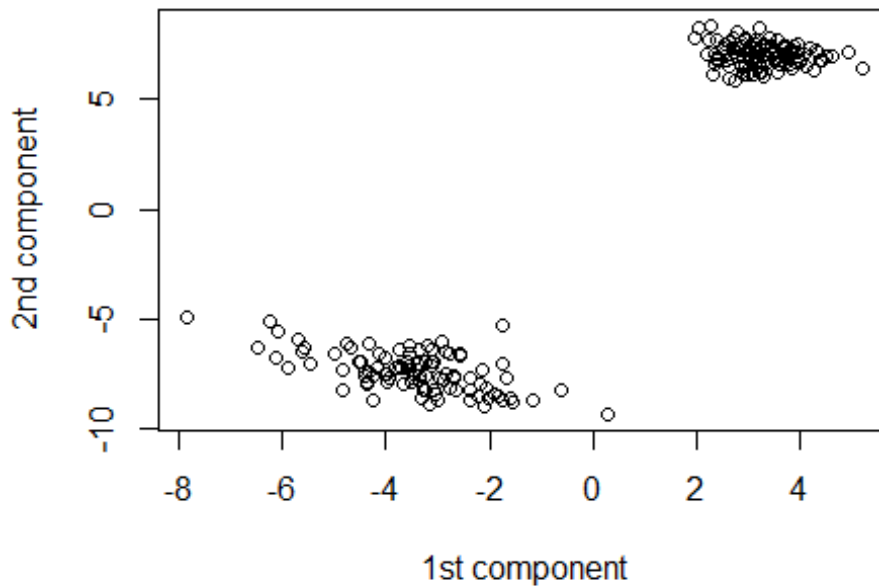


```

# plot best run representation
plot(non_params_mds_list[[best_run]]$points[, 1], non_params_mds_list[[best_run]]$points[, 2],
      xlab = "1st component", ylab = "2nd component", main = "Result of Best Non-metric MDS ")

```

Result of Best Non-metric MDS



```
# Metric MDS
metric_mds <- cmdscale(D, k = m)

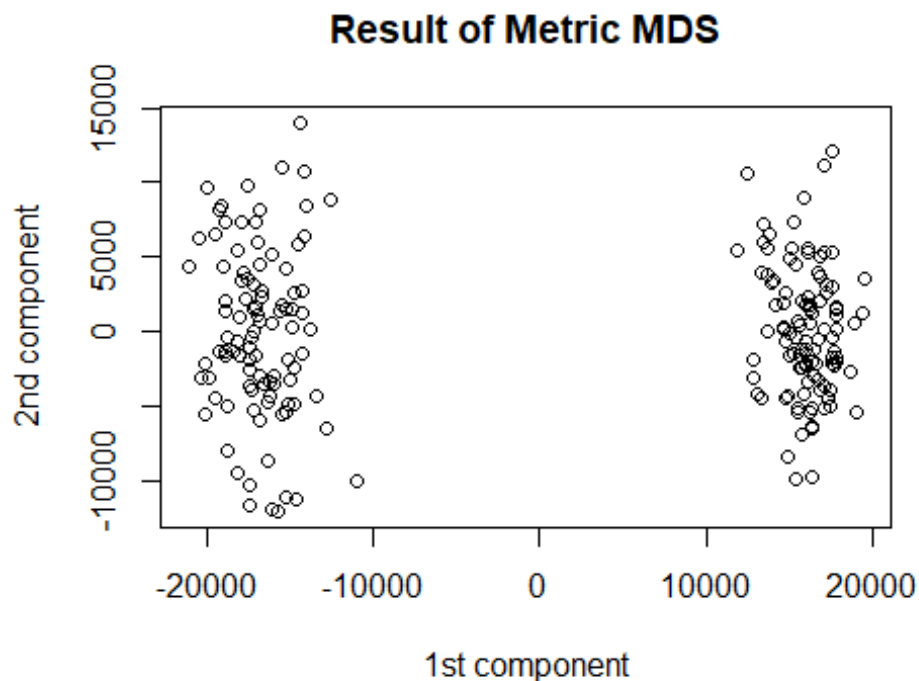
# Report stress of metric MDS
metric_stress <- sum((D - dist(metric_mds))^2)

## Warning in D - dist(metric_mds): la lunghezza più lunga dell'oggetto non è un
## multiplo della lunghezza più corta dell'oggetto

cat("Stress of Metric MDS: ", metric_stress, "\n")

## Stress of Metric MDS: 7.121805e+13

# Plot again maps for metric MDS for faster comparison
plot(metric_mds[, 1], metric_mds[, 2],
      xlab = "1st component", ylab = "2nd component", main = "Result of Metric MDS"
)
```



Metric and best non-metric mappings are strongly correlated. They both show that the dataset is heterogeneous. In particular, they both show there are 2 clusters well-divided. The main differences are the scale of the components in the different map, and the directions of maximum explained variance in the clusters of the 2 maps.

11

11. Compute the correlation matrix between the first two dimensions of the metric MDS and the two-dimensional solution of your best non-metric MDS. Comment your findings.

create a data frame to store the first two dimensions of metric and non-metric MDS

```
p <- data.frame(
  pc1_metric = metric_mds[, 1],
  pc2_metric = metric_mds[, 2],
  pc1_n_metric = non_params_mds_list[[best_run]]$points[, 1],
  pc2_n_metric = non_params_mds_list[[best_run]]$points[, 2]
)
```

#compute the correlation matrix

```
corr_matrix <- cor(p)
```

```
print(corr_matrix)
```

```
##          pc1_metric  pc2_metric pc1_n_metric pc2_n_metric
## pc1_metric  1.000000e+00 -6.702176e-17   0.95472789   0.99566188
## pc2_metric -6.702176e-17  1.000000e+00  -0.06874971   0.02506743
## pc1_n_metric 9.547279e-01 -6.874971e-02   1.00000000   0.93762408
## pc2_n_metric 9.956619e-01  2.506743e-02   0.93762408   1.00000000
```



```
cat("Correlation between pc1_metric and pc1_n_metric:", corr_matrix[1, 1], "\n")
## Correlation between pc1_metric and pc1_n_metric: 1
cat("Correlation between pc1_metric and pc2_n_metric:", corr_matrix[1, 2], "\n")
## Correlation between pc1_metric and pc2_n_metric: -6.702176e-17
cat("Correlation between pc2_metric and pc1_n_metric:", corr_matrix[2, 1], "\n")
## Correlation between pc2_metric and pc1_n_metric: -6.702176e-17
cat("Correlation between pc2_metric and pc2_n_metric:", corr_matrix[2, 2], "\n")
## Correlation between pc2_metric and pc2_n_metric: 1
```

Dimensions 1 and 2 of the metric MDS show a strong positive correlation with dimensions 1 and 2 of the non-metric MDS, respectively. The near-zero correlations with the cross-dimensions (pc1_metric vs pc2_n_metric and pc2_metric vs pc1_n_metric) suggest that these dimensions are relatively independent of each other.