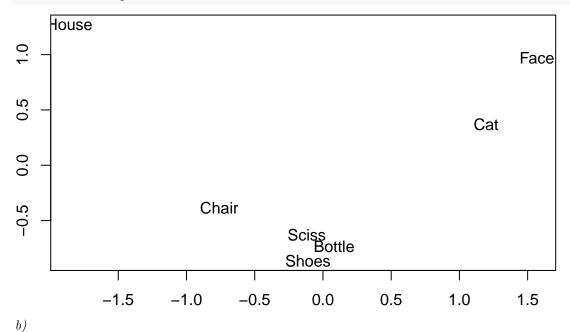
# 348HWw9

### Shay Lebovitz

## 11/8/2020

```
1) a)
brain <- read.table('/Users/shaylebovitz/R/Stat 348/brain.txt', header = TRUE)
brain <- as.dist(brain)
brain_mds <- cmdscale(brain, eig = TRUE)
plot(brain_mds$points, type = 'n', xlab = '', ylab = '')
text(brain_mds$points, labels = labels(brain))</pre>
```



brain\_mds\$GOF

#### ## [1] 0.7217058 0.7400254

Both methods for goodness of fit give results greater than 0.70, so I'd say the 2-D MDS model is fairly accurate.

c)

```
brain_mds$eig
```

```
## [1] 7.951622e+00 4.518817e+00 2.232862e+00 1.198964e+00 9.490998e-01 ## [6] 3.330669e-16 -4.277512e-01
```

We see that the first two eigengvectors are relatively large, but not that much different than the third. Overall, the MDS analysis is appropriate in this case with regards to the ability to distinguish between categories based on a brain scan.

```
2) a)
movies <- read.table('/Users/shaylebovitz/R/Stat 348/movies.txt', header = TRUE)
movies <- as.dist(movies)
movies_mds <- cmdscale(movies, eig = TRUE)
plot(movies_mds$points, type = 'n', xlab = '', ylab = '')
text(movies_mds$points, labels = labels(movies))</pre>
DKR
```

movies\_mds\$GOF

```
## [1] 0.7180194 0.8190584
```

```
movies_mds$eig
```

```
## [1] 6.594746e-01 4.080729e-01 1.686167e-01 6.721957e-02 -1.665335e-16 ## [6] -4.371786e-02 -1.396931e-01
```

We see that the first two eigenvalues are relatively large compared to the others, and the goodness-of-fits for each method are above 0.70, and thus we can say that the 2D MDS model is fairly accurate for the data.

- c) Because 'distance' in this case is a measure of the similarity in reviewers' ratings for a given movie (0 for all reviewers ratings are the same, 1 if all different), the axes in this case can be viewed as the variability, or perhaps the contentiousness/polarizability of a movie.
- d) Since she didn't like "Batman Forever", we would want to find the movie with the greatest distance away from it.

#### movies

```
##
           В
                BR
                      BF
                            BaR
                                   DK
                                          BB
## BR
       0.571
       0.250 0.500
  BF
   BaR 0.375 0.500 0.444
       0.500 0.500 0.778 0.778
  DK
       0.500 0.625 0.556 0.889 0.222
## DKR 0.667 0.667 0.857 0.857 0.143 0.857
```

That movie would be "Dark Knight Rises" (DKR, distance = 0.857). To find one she wouldn't like, we need

```
to find the most similar one. That would be "Batman" (B, distance = 0.250).
3) a)
colleges <- read.table('/Users/shaylebovitz/R/Stat 348/Colleges.txt', header = TRUE)</pre>
fnc <- function(x) {</pre>
  x/sd(x)
}
colleges <- apply(colleges, 2, fnc)</pre>
colleges <- as.data.frame(colleges)</pre>
b)
col_dist <- dist(colleges)</pre>
c)
colleges_mds <- cmdscale(col_dist, eig = TRUE)</pre>
plot(colleges_mds$points, type = 'n', xlab = '', ylab = '')
text(colleges_mds$points, labels = labels(col_dist))
                                             JIÇU
                                                        HU
                            SC
     NYU
1.0
                               GU <sup>UP</sup>
0.5
                                            NPO
                                                           SCI
                                                                            MIT
                                               YU
                     EU
                                            VU
                                 WU
                                COR
                                                                        CIT
                                           CMU
                                                  RU
                                          BU
                  BC
S
                                        ND DC
             -3
                       -2
                                                      1
                                                                2
                                                                          3
                                 -1
                                            0
d)
colleges mds$GOF
## [1] 0.6940049 0.6940049
colleges_mds$eig
    [1]
        5.794127e+01 2.533933e+01 1.943377e+01 1.169702e+01 5.588615e+00
##
         1.170022e-14 1.285613e-15 3.624999e-16 -5.522422e-17 -1.320100e-16
## [11] -1.983912e-16 -3.222912e-16 -3.585603e-16 -5.187469e-16 -8.359932e-16
## [16] -9.099191e-16 -1.473222e-15 -1.519925e-15 -1.605773e-15 -1.825994e-15
## [21] -1.884079e-15 -2.749155e-15 -2.967755e-15 -3.156702e-15 -5.411358e-15
```

The goodness of fit by both methods is 0.69, which is right below the usual lowest accepted value of 0.70. I think it would still be fine, as 0.70 is somewhat arbitrarily chosen. We see that the third eigenvalue is relatively similar in magnitude to the second, so perhaps a 3D MDS representation would be more appropriate.

e) It is hard to tell exactly from the distance matrix, but from the plot it looks like the four most similar colleges would be Duke (DU), Penn (UP), Yale (YU), and either Georgetown (GU), U.Chicago (UC), or Stanford (SU). The least similar would be Notre Dame (ND), Dartmouth (DC), Princeton (PU), and Boston College (BC).