

Psychology 594

Final Take Home Exercise; Fall 2013

November 12, 2013

Questions I and II are to be done with your chosen proximity matrix (from Michael Lee's web site; remember the adolescent drug use data are "off limits" as are the number data).

http://cda.psych.uiuc.edu/multivariate_class_final_2013.zip

I. (a) From routines available in the Matlab Statistical Toolbox, carry out a complete-link hierarchical clustering and interpret your results. Use at least the following three M-files:

`squareform.m`
`linkage.m`
`dendrogram.m`

(b) From the Cluster Analysis Toolbox M-files:

(i) use `order.m`; `ultrafit.m` (with a complete-link target); `ultrafnd.m` (with `randperm` several times).

(ii) use `ultrafnd_confit.m` (with the order found in (i)),
and `ultrafnd_confnd.m`.

(iii) use `partitionfnd_averages.m`; `partitionfnd_diameters.m`;
and then `partitionfit.m` after each.

(iv) use `cent_ultrafnd_confit.m` (with the order found in (i)), and
`cent_ultrafnd_confnd.m`.

(v) use `atreefit.m` (with a complete-link target); `atreefnd.m` (with `randperm` several times); `atreedec.m`, and `ultraorder.m`.

(vi) use `consec_subsetfit.m` and `consec_subsetfit_alter.m`.

Again, interpret the results obtained from the various analyses.

II. Carry out a multiple restart Monte Carlo on your data using an M-script parallel to `ms_script_yourdata_mds.m`. Interpret the results obtained in relation to the previous cluster analyses in I. Label the various graphs with your particular data matrix title and your name. If the nonmetric option causes numerical difficulties, follow one of the two options mentioned in class: first, cut the iterations from 100 to 10; and if that doesn't help, change '`stress`' to '`sstress`'.

```
load decathlon.dat
```

```
decathlon_dissimilarities = 1 - decathlon;
```

```
decathlon_dissimilarities
```

```
n = 10;
```

```
tic;
```

```
opts = statset('Maxiter',1000);
```

```
best_vaf = 0.0;
```

```
store_vaf = zeros(100,1);
```

```
for k = 1:100
```

```

[coords,stress] = ...
mdscale(decathlon_dissimilarities,2,'Criterion',...
'metricstress','Start','random',...
'Replicates',1,'Options',opts);

n = size(coords,1);

distance_matrix = zeros(n,n);

for i = 1:n
    for j = 1:n

        distance_matrix(i,j) = ...
            sqrt(((coords(i,1) - coords(j,1))^2) + ...
                ((coords(i,2) - coords(j,2))^2));
    end
end

decathlon_vec = squareform(decathlon_dissimilarities);

distance_vec = squareform(distance_matrix);

r = corrcoef(decathlon_vec',distance_vec');

vaf = r(1,2)^2;

store_vaf(k) = vaf;

if(vaf > best_vaf)

    best_vaf = vaf;

```

```

        best_coords = coords;
        best_distance_vec = distance_vec;

end
end

sorted_vafs = sort(store_vaf');

sorted_vafs
best_vaf
best_coords
best_distance_vec

figure(1)

axis equal

plot(best_coords(:,1),best_coords(:,2),'ko')

hold on

for i = 1:n

    objectlabels{i,1} = int2str(i);

end

text(best_coords(:,1),best_coords(:,2),objectlabels,...
'fontsize',10,'verticalalignment','bottom')

toc;

```

```

euclidean_coordinates = [best_coords(:,1),best_coords(:,2)];

figure(2)

axis equal

plot(decathlon_vec,best_distance_vec,'bo')

hold on

xlabel('Dissimilarities')
ylabel('Distances')

tic;

best_vaf = 0.0;

store_vaf = zeros(100,1);

best_disparities = zeros(n,n);

for k = 1:100

[coords,stress,disparities] = ...
mdscale(decathlon_dissimilarities,2,'Criterion',...
'stress','Start',...
'random','Replicates',1,'Options',opts);

n = size(coords,1);

distance_matrix = zeros(n,n);

```

```

for i = 1:n
    for j = 1:n

        distance_matrix(i,j) = ...
            sqrt(((coords(i,1) - coords(j,1))^2) + ...
                ((coords(i,2) - coords(j,2))^2));
    end
end

decathlon_vec = squareform(decathlon_dissimilarities);

distance_vec = squareform(distance_matrix);

r = corrcoef(decathlon_vec',distance_vec');

vaf = r(1,2)^2;

store_vaf(k) = vaf;

if(vaf > best_vaf)

    best_vaf = vaf;
    best_coords = coords;
    best_disparities = disparities;
    best_distance_vec = distance_vec;

end end

store_vaf;

```

```

sorted_vafs = sort(store_vaf');

sorted_vafs
best_vaf
best_coords

figure(3)

axis equal

plot(best_coords(:,1),best_coords(:,2),'ko')

hold on

for i = 1:n

    objectlabels{i,1} = int2str(i);

end

text(best_coords(:,1),best_coords(:,2),objectlabels,...
'fontsize',10,'verticalalignment','bottom')

toc;

euclidean_coordinates_nonmetric = ...
[best_coords(:,1),best_coords(:,2)];

best_disparities_vec = squareform(best_disparities);

best_distance_vec

```

```

best_disparities_vec

figure(4)

axis equal

[dum,ord] = sortrows([best_disparities_vec(:) decathlon_vec(:)]);

plot(decathlon_vec,best_distance_vec,'bo',...
decathlon_vec(ord),best_disparities_vec(ord),'r.-')

hold on

xlabel('Dissimilarities')
ylabel('Distance/Disparities')

legend({'Distances' 'Disparities'}, 'Location', 'NW')


[d,z,transform] = ...
procrustes(euclidean_coordinates_nonmetric,euclidean_coordinates);

figure(5)

axis equal

plot(euclidean_coordinates_nonmetric(:,1),...
euclidean_coordinates_nonmetric(:,2),'rx',...
euclidean_coordinates(:,1),...
euclidean_coordinates(:,2),'b.',...
```



```

z(:,1),z(:,2),'ko')

hold on

text(euclidean_coordinates_nonmetric(:,1),...
euclidean_coordinates_nonmetric(:,2),objectlabels,...
'fontsize',8,'verticalalignment','bottom')

text(z(:,1),z(:,2),objectlabels,'fontsize',8,...
'verticalalignment','bottom')

transform(1).b

transform(1).T

transform(1).c

```

III. The data matrices

`supreme_court_11_12.dat` and `supreme_court_12_13` give the proportions of cases that a given pair of Supreme Court justices *disagreed* on during the 2011/2012 and 2012/2013 court terms. Thus, the numbers can be treated as dissimilarities. The order of the rows and columns is as follows (for both matrices):

- 1: Roberts
- 2: Scalia
- 3: Kennedy
- 4: Thomas
- 5: Ginsburg
- 6: Breyer
- 7: Alito
- 8: Sotomayor
- 9: Kagan

Using the M-files, `order.m`, `linfitac.m`, and `ultrafnd.m`, evaluate whether a unidimensional scaling (i.e., a “continuous” model) or an ultrametric (a “categorical” model) gives a better fit for both matrices. Interpret the results of your analyses in terms of the political composition of the court in the 11/12 and 12/13 terms. Does anyone stand out as a “swing vote”? How is this shown in the unidimensional scaling and in the hierarchical clustering? If you wish some background reading, see Adam Liptak, *In Supreme Court Term, Striking Unity on Major Cases* (*New York Times*, June 30, 2012), and *Roberts Pulls Supreme Court to the Right Step by Step* (*New York Times*), June 27, 2013).

Finally, using the dendrogram command in Matlab and the `'Reorder'` option, display the dendrogram using the same orders you found using `order.m`. Are there difficulties with any of the branches now crossing?

The contents of the file `supreme_court_11_12.dat`:

```

0 14 16 12 36 30 9 29 27
14 0 24 7 44 43 12 36 34
16 24 0 19 27 23 22 22 17
12 7 19 0 44 37 12 36 31
36 44 27 44 0 17 43 16 15
30 43 23 37 17 0 33 20 15
9 12 22 12 43 33 0 36 30
29 36 22 36 16 20 36 0 16
27 34 17 31 15 15 30 16 0

```

The contents of the file `supreme_court_12_13.dat`:

```

0 15 15 14 35 26 10 35 33
15 0 23 14 32 38 23 35 32
15 23 0 24 27 23 19 25 28
14 14 24 0 41 32 12 39 39

```

35	32	27	41	0	12	42	6	4
26	38	23	32	12	0	30	13	9
10	23	19	12	42	30	0	39	41
35	35	25	39	6	13	39	0	4
33	32	28	39	4	9	41	4	0