

Psychology 594

Final Take Home Exercise

November 18, 2009

Questions I and II are to be done with your chosen proximity matrix (from Michael Lee's web site).

http://cda.psych.uiuc.edu/multivariate_class_final_2009.zip

I. (a) From routines available in the Matlab Statistical Toolbox, carry out a complete-link hierarchical clustering. 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`.

II. Carry out multiple restart Monte Carlo on your data using an M-script parallel to `ms_script_yourdata_mds.m`.

```
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',...
'metricsstress','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',...
'sstress','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 matrix `supreme_court_08_09.dat` gives the number of (non-unanimous) cases (out of 53) that a given pair of Supreme Court justices *disagreed* on during the 08/09 court term. Thus, the numbers can be treated as dissimilarities. The order of the rows and columns is as follows:

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

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. Interpret the results of your analyses in terms of the political composition of the court in the 08/09 term. If you wish some background reading, see Adam Liptak, *Roberts Court Shifts Right, Tipped by Kennedy* (*New York Times*, July 1, 2009). The contents of the file `supreme_court_08_09.dat`:

```
0 11 15 11 26 37 37 35 37
11 0 18 10 27 34 39 32 34
15 18 0 15 18 27 27 31 33
11 10 15 0 32 39 41 41 43
26 27 18 32 0 11 10 15 16
37 34 27 39 11 0 6 10 14
37 39 27 41 10 6 0 10 12
35 32 31 41 15 10 10 0 10
37 34 33 43 16 14 12 10 0
```