Project 2 – Unsupervised Learning – MVA – ENS Paris-Saclay

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Due Date: 11/24/2017

INSTRUCTIONS: This project is to be done in MATLAB or Phyton and in groups of three students. Please report which student did which part. Each student should also email TA a grade (0-100%=best) for the work of his/her mates.

 Clustering Algorithms. Implement the Spectral Clustering (SC) algorithm (Algorithm 4.5), K-Subspaces algorithm (Algorithm 6.1) and Sparse Subspace Clustering (SSC) algorithm with noisy data (Algorithm 8.5 and 8.6). You can use the MATLAB kmeans function (or python) inside SC.

```
1 function groups = SpectralClustering(Affinity, n)
2 % Affinity: N by N affinity matrix, where N is the number of points.
3 % n: number of groups
```

```
1 function [global_groups, global_obj] = ksubspaces(data, n, d, replicates)
2 % data: D by N data matrix.
3 % n: number of subspaces
4 % d: dimension of subspaces
5 % replicates: number of restarts
```

```
1 function groups = SSC(data, n, tau, mu2)
2 % data: D by N data matrix.
3 % n: number of clusters
4 % tau, mu2: parameter
```

Implement also a function for computing the clustering error. Note that the true labels for 9 points in two subspaces may be $[1 \ 1 \ 1 \ 1 \ 2 \ 2 \ 2 \ 2 \ 2]$ while an algorithm may produce $[2 \ 2 \ 2 \ 2 \ 1 \ 1 \ 1 \ 1]$, which is a correct clustering. Therefore, you need to evaluate the error for all possible permutations of n letters, where n is the number of groups. When n is large, this can be time consuming. You may use an existing implementation of the Hungarian algorithm to do this efficiently.

```
1 error = clustering_error(label, groups)
2 % label: N-dimensional vector with ground truth labels for a dataset with N points
3 % groups: N-dimensional vector with estimated labels for a dataset with N points
```

2. Face Clustering. Apply the three algorithms in part (a) to the face images from the ExtendedYaleB dataset. The data is already in matrix format. The file ExtendedYaleB.mat contains two variables EYALEB_DATA and EYALEB_LABEL. EYALEB_DATA is a data matrix containing face images of 38 subjects, each under 64 different illumination conditions. Each column contains a face image of size 48*42, which has been flattened to a vector of size 2016. Use the following command to show a face image:

reshape(uint8(EYALEB_DATA(:, 1)), [48, 42]))

EYALEB_LABEL is a vector where each entry is the label (ranging from 1 to 38) for the face in the corresponding column of EYALEB_DATA. Report the clustering error for individuals 1-2 for different choices of the parameters σ and K for SC (when constructing a Gaussian affinity with K-NN), number of restarts for K-subspaces, and λ for SSC. Once you have chosen the best parameters for each method, report the clustering errors for individuals 1-2, 1-10, and 1-20, 1-30, 1-38, and analyze the clustering error as a function of the number of groups.

3. **Motion Segmentation.** Apply the algorithms in part (a) to the feature point trajectories from the Hopkins155 dataset. You may use the code below to extract the data matrix.

```
d = dir;
1
2
  for i = 1:length(d)
       if ( (d(i).isdir == 1) && ¬strcmp(d(i).name,'.') && ¬strcmp(d(i).name,'..') )
3
4
            filepath = d(i).name;
            eval(['cd ' filepath]);
5
6
7
            f = dir;
            foundValidData = false;
8
            for j = 1:length(f)
9
                if ( ¬isempty(strfind(f(j).name, '_truth.mat')) )
10
                    ind = j;
11
                     foundValidData = true;
12
13
                    break
14
                end
15
            end
            eval(['load ' f(ind).name]);
16
17
            cd ..
18
19
            if (foundValidData)
                n = max(s);
20
                N = size(x, 2);
21
22
                F = size(x, 3);
                D = 2 \star F;
23
                X = reshape(permute(x(1:2,:,:),[1 3 2]),D,N);
24
25
            end
26
  end
```

Report the clustering error for different choices of the parameters σ and K for SC (when constructing a Gaussian affinity with K-NN), number of restarts for K-subspaces, and λ for SSC. Compare the best results.

Submission instructions. Please send an email to the TA with subject MVA-Supervised-Learning-Project2 and attachment lastname1-lastname2-lastname3-MVA-Supervised-Learning-Project2.zip or lastname1-lastname2-lastname3-MVA-Supervised-Learning-Project2.tar.gz. The attachment should contain a file called README, which contains instructions on how to run your code. The TA will run your scripts to generate the results. Please include a report in PDF format containing figures and tables with your results as well as an explanation of what you did an analysis of the results.