Homework 6

Please scan and upload your assignments on or before April 30, 2020.

- You are encouraged to discuss ideas with each other; but
- you must acknowledge your collaborator, and
- you must compose your own writeup and/or code independently.
- We **strongly** encourage answers to theory questions in Latex, and answers to coding questions in Python (Jupyter notebooks).
- Please upload your solutions in the form of a single .pdf or .zip file on NYUClasses.
- Maximum score: 50 points.
- 1. (10 points) Assume that you have 4 samples each with dimension 3, described in the data matrix X,

$$X = \left[\begin{array}{rrr} 3 & 2 & 1 \\ 2 & 4 & 5 \\ 1 & 2 & 3 \\ 0 & 2 & 5 \end{array} \right]$$

For the problems below, you may do the calculations in python (or R or Matlab). Explain your calculations in each step.

- a. Find the sample mean.
- Zero-center the samples, and find the eigenvalues and eigenvectors of the data covariance matrix Q.
- c. Find the PCA coefficients corresponding to each of the samples in X.
- d. Reconstruct the original samples from the top two principal components, and report the reconstruction error for each of the samples.
- 2. (10 points) In class, we analyzed the per-iteration complexity of k-means. Here, we will prove that the k-means algorithm will terminate in a finite number of iterations. Consider a data set $X = \{x_1, \ldots, x_n\} \in \mathbb{R}^d$.
 - a. Show that the k-means loss function can be re-written in the form:

$$F(\eta, \mu) = \sum_{i=1}^{n} \sum_{j=1}^{k} \eta_{ij} ||x_i - \mu_j||^2$$

where $\eta = (\eta_{ij})$ is a suitable binary matrix (with 0/1 entries). Provide a precise interpretation of η .

b. Show that each iteration of Lloyd's algorithm can only decrease the value of F.

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c. Conclude that the algorithm will terminate in no more than T iterations, where T is some finite number. Give an upper bound on T in terms of the number of points n.

- 3. (10 points) Using the Senate Votes dataset demo'ed in Lecture 11, perform k-means clustering with k=2 and show that you can learn (most of) the Senators' parties in a completely unsupervised manner. Which Senators did your algorithm make a mistake on, and why?
- 4. (20 points) The *Places Rated Almanac*, written by Boyer and Savageau, rates the livability of several US cities according to nine factors: climate, housing, healthcare, crime, transportation, education, arts, recreation, and economic welfare. The ratings are available in tabular form, available as a supplemental text file. Except for housing and crime, higher ratings indicate better quality of life. Let us use PCA to interpret this data better.
 - a. Load the data and construct a table with 9 columns containing the numerical ratings. (Ignore the last 5 columns they consist auxiliary information such as longitude/latitude, state, etc.)
 - b. Replace each value in the matrix by its base-10 logarithm. (This pre-processing is done for convenience since the numerical range of the ratings is large.) You should now have a data matrix X whose rows are 9-dimensional vectors representing the different cities.
 - c. Perform PCA on the data. Remember to center the data points first by computing the mean data vector μ and subtracting it from every point. With the centered data matrix, do an SVD and compute the principal components.
 - d. Write down the first two principal components v_1 and v_2 . Provide a qualitative interpretation of the components. Which among the nine factors do they appear to correlate the most with?
 - e. Project the data points onto the first two principal components. (That is, compute the highest 2 scores of each of the data points.) Plot the scores as a 2D scatter plot. Which cities correspond to outliers in this scatter plot?
 - f. Repeat Steps 2-5, but with a slightly different data matrix instead of computing the base-10 logarithm, use the z-scores. (The z-score is calculated by computing the mean μ and standard deviation σ for each feature, and normalizing each entry x by $\frac{x-\mu}{\sigma}$). How do your answers change?