Course 20874, Assignment 1, Problem 2

Giacomo Cirò

1 Theoretical Model

As shown in class, the optimal centroid for a given cluster of points according to the squared L2-norm is the mean of the points belonging to that cluster. Hence, the possible centroids given a set of points are a finite number, i.e., the means of all possible subsets of points.

Notice that multiple occurences of the same point must be assigned to the same centroid. Nonetheless, these multiple occurences are important as they skew the mean of the cluster they belong to and magnify the overall loss. Aware of this, we can simplify this problem by considering only unique points and use their relative frequency as a weight when computing tentative centroids and their absolute frequency as weight when computing the objective function. Formally, we model the problem as follows:

Given a set of P points, denote with i = 1, ..., N the unique points with coordinates $p_i \in \mathbb{R}^n$, clearly $N \leq P$. Denote with $j = 1, ..., 2^N - 1$ the possible centroids with coordinates c_j , obtained as follows:

$$c_j = \frac{1}{\sum_{i \in S_j} w_i} \sum_{i \in S_i} p_i \cdot w_i$$

where S_j are all possible subsets of unique points and w_i their frequency in the original set. We also define the total cost of cluster S_j as:

$$d_j = \sum_{p_i \in S_j} w_i ||c_j - p_i||^2$$

and define the cluster matrix X as:

$$X_{ij} = \mathbb{1}_{S_i}(i)$$

Where \mathbb{F}_{S_j} is the indicator function. We also define a binary variable $y_j \in \{0,1\}$ indicating whether or not we choose centroid j and the corresponding cluster S_j .

We now model K-means as follows:

$$\min_{y_j} \sum_{j=1}^{2^{N}-1} y_j d_j$$
s.t.
$$\sum_{j=1}^{s^{N}-1} y_j X_{ij} = 1 \quad \forall i = 1, ..., N$$

$$\sum_{j=1}^{s^{N}-1} y_j = k$$

The constraints force to assign each point to exactly 1 centroid and to choose at most k centroids. To optimize running time, we can perform the following semplification and avoid considering all the $2^N - 1$ possible clusters:

- discard all clusters such that the squared distance with respect to the generated centroid is greater than the objective function value found with K-means
- discard all clusters with more than N-k+2 elements. In fact, choosing one such cluster would lead to k-2 points left, but only k-1 clusters to be generated. It's now optimal to cluster k-2 points on their own. This leaves one free cluster yet to be assigned. It's always optimal to remove one point from the first cluster and assign it to this last one.

2 Implementation

In order to compress an image using IP run the following commands:

- 1. python3 recolor.py 20col.png 8col.png 8 to get the compression using Lloyd algorithm;
- 2. python3 generate_datafile.py to generate the centroids and write the .dat file (either an integer is passed as argument or the program automatically uses the loss from the previously generated image as threshold when evaluating tentative clusters);
- 3. glpsol --model main.mod --data main.dat > out.txt to solve the IP and save the result;
- 4. python3 visualize_compressions.py to visualize the different compressions (IP-compressed image is generated by parsing out.txt).