

Homework 4

Out: *Nov 13*Due: *Dec 2*

1. Implement the portfolio management appearing in the notes for Lecture 16 in any programming environment and check its performance on S&P stock data (download from <http://ocobook.cs.princeton.edu/links.htm>). Include your code as well as the final performance (i.e., the percentage gain achieved by your strategy).
2. Consider a set of n objects (images, songs etc.) and suppose somebody has designed a *distance* function $d(\cdot)$ among them where $d(i, j)$ is the distance between objects i and j . We are trying to find a geometric realization of these distances. Of course, exact realization may be impossible and we are willing to tolerate a factor 2 approximation. We want n vectors u_1, u_2, \dots, u_n such that $d(i, j) \leq \|u_i - u_j\|_2 \leq 2d(i, j)$ for all pairs i, j . Describe a polynomial-time algorithm that determines whether such u_i 's exist.
3. The course webpage links to a grayscale photo. Interpret it as an $n \times m$ matrix and run SVD on it. What is the value of k such that a rank k approximation gives a reasonable approximation (visually) to the image? What value of k gives an approximation that looks high quality to your eyes? Attach both pictures and your code. (In matlab you need `mat2gray` function.) Extra credit: Try to explain from first principles why SVD works for image compression at all.
4. Suppose we have a set of n images and for some multiset E of image pairs we have been told whether they are *similar* (denoted +edges in E) or *dissimilar* (denoted -edges). These ratings were generated by different users and may not be mutually consistent (in fact the same pair may be rated as + as well as -). We wish to *partition* them into clusters S_1, S_2, S_3, \dots so as to maximise:

$$(\# \text{ of +edges that lie within clusters}) + (\# \text{ of -edges that lie between clusters}).$$

Show that the following SDP is an upperbound on this, where $w^+(ij)$ and $w^-(ij)$ are the number of times pair i, j has been rated + and - respectively.

$$\begin{aligned} \max \quad & \sum_{(i,j) \in E} w^+(ij)(x_i \cdot x_j) + w^-(ij)(1 - x_i \cdot x_j) \\ & |x_i|_2^2 = 1 \quad \forall i \\ & x_i \cdot x_j \geq 0 \quad \forall i \neq j. \end{aligned}$$

5. For the problem in the previous question describe a clustering into 4 clusters that achieves an objective value 0.75 times the SDP value. (Hint: Use Goemans-Williamson style rounding but with two random hyperplanes instead of one. You may need a quick matlab calculation just like GW.)

6. Suppose you are given m halfspaces in \mathfrak{R}^n with rational coefficients. Describe a polynomial-time algorithm to find the largest *sphere* that is contained inside the polyhedron defined by these halfspaces.
7. Let f be an n -variate convex function such that for every x , every eigenvalue of $\nabla^2 f(x)$ lies in $[m, M]$. Show that the optimum value of f is lowerbounded by $f(x) - \frac{1}{2m} \|\nabla f(x)\|_2^2$ and upperbounded by $f(x) + \frac{1}{2M} \|\nabla f(x)\|_2^2$, where x is any point. In other words, if the gradient at x is small, then the value of f at x is near-optimal. (Hint: By the mean value theorem, $f(y) = f(x) + \nabla f(x)^T(y - x) + \frac{1}{2}(y - x)^T \nabla^2 f(z)(y - x)$, where z is some point on the line segment joining x, y .)