

# Linear Algebra for Electrical Engineers

## Homework VI

Due 11/03 23:59

Please checkout the eTL homework announcement for the submission format. Skeleton code is available at this link: <https://github.com/3dvision-snu/linear-algebra-2020-fall>

1. The 2-vectors  $\mathbf{p}_1, \dots, \mathbf{p}_N$  represent the locations or positions of  $N$  objects, for example, factories, warehouses, and stores. The last  $K$  of these locations are fixed and given; the goal in a *placement problem* is to choose the locations of the first  $N - K$  objects. Our choice of the locations is guided by an undirected graph; an edge between two objects means we would like them to be close to each other. In *least squares placement*, we choose the locations  $\mathbf{p}_1, \dots, \mathbf{p}_{N-K}$  so as to minimize the sum of the squares of the distances between objects connected by an edge,

$$\|\mathbf{p}_{i_1} - \mathbf{p}_{j_1}\|^2 + \dots + \|\mathbf{p}_{i_L} - \mathbf{p}_{j_L}\|^2,$$

where the  $L$  edges of the graph are given by  $(i_1, j_1), \dots, (i_L, j_L)$ .

- (a) Let  $\mathcal{D}$  be the Dirichlet energy of the graph, as defined on page 135 of the textbook. Show that the sum of squared distance between the  $N$  objects can be expressed as  $\mathcal{D}(\mathbf{u}) + \mathcal{D}(\mathbf{v})$ , where  $\mathbf{u} = ((\mathbf{p}_1)_1, \dots, (\mathbf{p}_N)_1)$  and  $\mathbf{v} = ((\mathbf{p}_1)_2, \dots, (\mathbf{p}_N)_2)$  are  $N$ -vectors containing the first and second coordinates of the objects, respectively. (0.1 points)
- (b) Express the least squares placement problem as a least squares problem, with variable  $\mathbf{x} = (\mathbf{u}_{1:(N-K)}, \mathbf{v}_{1:(N-K)})$ . In other words, express the objective above (the sum of squares of the distances across edges) as  $\|A\mathbf{x} - \mathbf{b}\|$ , for an appropriate  $m \times n$  matrix  $A$  and  $m$ -vector  $\mathbf{b}$ . You will find that  $m = 2L$ . *Hint.* Recall that  $\mathcal{D}(\mathbf{y}) = \|B^\top \mathbf{y}\|^2$ , where  $B$  is the incidence matrix of the graph. (0.1 points)
- (c) Solve the least squares placement problem with numpy for the specific problem with  $N = 10, K = 4, L = 13$ , fixed locations

$$\mathbf{p}_7 = (0, 0), \mathbf{p}_8 = (0, 1), \mathbf{p}_9 = (1, 1), \mathbf{p}_{10} = (1, 0)$$

and edges  $(1, 3), (1, 4), (1, 7), (2, 3), (2, 5), (2, 8), (2, 9), (3, 4), (3, 5), (4, 6), (5, 6), (6, 9), (6, 10)$ . Write the final results to the report. (0.5 points)

2. Suppose that  $A$  has linearly independent columns, with  $\hat{\mathbf{x}} = A^\dagger \mathbf{b}$  minimizing  $\|A\mathbf{x} - \mathbf{b}\|^2$ . In this exercise, you are to implement an iterative method, named *Richardson iteration* to compute least squares solution  $\hat{\mathbf{x}}$ . We randomly initialize  $\mathbf{x}^{(0)}$  and for  $k = 0, 1, \dots$ , and iterate

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - \frac{1}{\|A\|} A^\top (A\mathbf{x}^{(k)} - \mathbf{b}),$$

This iteration is proven to converge to least square solution  $\hat{\mathbf{x}}$ . Generate a random  $20 \times 10$  matrix  $A$  and 20-vector  $\mathbf{b}$  and compute  $\hat{\mathbf{x}} = A^\dagger \mathbf{b}$ . Run the algorithm for 300 iterations and plot  $\|\mathbf{x}^{(k)} - \hat{\mathbf{x}}\|$  to verify the convergence. You may use `np.linalg.norm` to compute the norm of matrices and vectors. Attach the plot to the report. (0.3 points)