## Exercise 8

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## 1 Laplacian Smoothing

## 1.1 (A)

The laplacian L = W - A can be decomposed into a diagonal matrix W and a off-diagonal matrix -A. Therefore, we only have to show that the diagonal elements of  $D^TD$  form W and the off-diagonal elements of  $D^TD$  form -A.

For the diagonal elements,  $L_{ii} = D_i \cdot D_i$  where  $D_i$  is the *i*-th column of the D matrix and  $\cdot$  denotes inner product. From the definition of the oriented edge matrix D we know that  $D_{ki} = -1$  if there is an edge k whose ending node is i and  $D_{ki} = 1$  if edge k's beginning node is i. Therefore,  $L_{ii} = D_i \cdot D_i$  is just the degree of node i. That is,  $L_{ii} = W_i$ .

For the off-diagonal elements,  $L_{ij} = D_i \cdot D_j$ . We only need to examine the indices k's for which  $D_{ki} \neq 0$  and  $D_{kj} \neq 0$ . This only happens when edge k connects node i and node j, in which case  $D_{ki}D_{kj} = -1$ . This implies that  $L_{ij} = -Aij$  for  $i \neq j$ .

## 1.2 (B)

Given the Laplacian smoothing problem:

$$\underset{x \in \mathbb{R}^n}{\text{minimize}} \quad \frac{1}{2} \|y - x\|_2^2 + \frac{\lambda}{2} x^T L x, \tag{1}$$

we first show that it is convex.

The first term is a norm, which is always convex. The second term is convex iff L is positive semidefinite. This is true because L is diagonally dominant and symmetric with nonnegative diagonal elements.

With convexity, we can safely invoke the first-order optimality condition, which leads to:

$$(\lambda D^T D + I)\hat{x} = y. (2)$$