David J. Elkind - LIPO

I'm interested in estimating the Lipschitz constant k for functions which approximate as Lipschitz. When the functions are noisy, they can't be Lipschitz, but adding a noise term s can make the functions behave more nicely; including the large penalty 10^6 for s forces it to often be 0.

The function of interest is f and it is only observed via some finite number of points $(x_i, f(x_i))$ for i = 1, 2, ...t. Define

$$U(y) = \min_{i \in 1, 2, \dots t} \left[f(x_i) + \sqrt{s_i + (y - x_i)^\top K(y - x_i)} \right]$$
 (1)

One way to estimate k is to optimize the following program (using a different k_i for each coordinate):

$$\min_{K,s} ||K||_F^2 + 10^6 \sum_{i=1}^t s_i^2 \tag{2}$$

s.t.
$$U(x_i) \ge f(x_i)i \in \{1, 2, \dots t\}$$
 (3)

$$s_i \ge 0 \forall i \in \{1, 2, \dots t\} \tag{4}$$

$$K_{i,j} \ge 0 \forall i, j \in \{1, 2, \dots d\}^2$$
 (5)

Expressing a signle constraint, we have

$$U(x_i) \ge f(x_i) \ \forall i = 1, 2, \dots, t \tag{7}$$

$$\implies f(x_j) + \sqrt{s_j + (x_i - x_j)^\top K(x_i - X_j)} \ge f(x_i)$$
(8)

$$s_i + (x_i - x_i)^\top K(x_i - x_i) \ge (f(x_i) - f(x_i))^2$$
 (9)

(10)

Which is linear in s and K. Moreover, this is exactly the definition of Lipschitz continuity when $s_i = 0$.

This is a little bit of a bummer, though, since this isn't symmetric in i, j due to the presence of the s_j . This implies that there are $t^2 > {t \choose 2}$ constraints, which might be too many for large problems.