

# Formulating the optimization problem

## The optimization problem

**Problem:** Given two linearly separable sets of points  $A^\pm \subset \mathbb{R}^k$ :

$$A^+ = \{x_1^+, \dots, x_{n_+}^+\}$$

and

$$A^- = \{x_1^-, \dots, x_{n_-}^-\}$$

Find points  $p \in C(A^+)$  and  $q \in C(A^-)$  so that

$$\|p - q\| = \min_{\substack{p' \in C(A^+) \\ q' \in C(A^-)}} \|p' - q'\|$$

## The optimization problem (continued)

As above, let our linearly separable sets be

$$A^+ = \{x_1^+, \dots, x_{n_+}\}$$

and

$$A^- = \{x_1^-, \dots, x_{n_-}\}$$

**Problem 1:** Let  $\lambda^\pm = (\lambda_1^\pm, \dots, \lambda_{n_\pm}^\pm)$  be two vectors of real numbers of length  $n_\pm$  respectively. Define

$$w(\lambda^+, \lambda^-) = \sum_{i=1}^{n_+} \lambda_i^+ x_i^+ - \sum_{i=1}^{n_-} \lambda_i^- x_i^-.$$

Find  $\lambda^\pm$  such that  $\|w(\lambda^+, \lambda^-)\|$  is minimal subject to the conditions that all  $\lambda_i^\pm \geq 0$  and  $\sum_{i=1}^{n_\pm} \lambda_i^\pm = 1$ .

## The optimization problem (continued)

Notice that:

- $w(\lambda^+, \lambda^-)$  is a quadratic function in the  $\lambda$ 's with coefficients coming from the dot products of the  $x_i^\pm$ .
- The constraints are *inequalities* rather than equalities, so a direct application of Lagrange multipliers as we learned in calculus won't work.
- We will describe an iterative algorithm for solving this problem called Sequential Minimal Optimization, due to John Platt and introduced in 1998.
- First, though, we reformulate the problem slightly.

## Reformulating the constrained optimization problem.

**Problem 2:** Let

$$Q(\lambda^+, \lambda^-) = \|w(\lambda^+, \lambda^-)\|^2 - \sum_{i=1}^{n_+} \lambda_i^+ - \sum_{i=1}^{n_-} \lambda_i^-.$$

Let  $\lambda^\pm$  be values that minimize  $Q(\lambda^+, \lambda^-)$  where all  $\lambda_i^\pm \geq 0$  and

$$\alpha = \sum_{i=1}^{n_+} \lambda_i^+ = \sum_{i=1}^{n_-} \lambda_i^- > 0.$$

Then  $\tau^\pm = (1/\alpha)\lambda^\pm$  is a solution to optimization problem 1.

## Equivalence of the reformulated problem

**Proof:** We have  $(\lambda^+, \lambda^-)$  solving problem 2 and  $(\sigma^+, \sigma^-)$  solving problem 1; and finally we have  $(\tau^+, \tau^-) = (1/\alpha)(\lambda^+, \lambda^-)$ .

1. Since  $(\tau^+, \tau^-)$  satisfy the constraints of problem 2, we have:

$$Q(\lambda^+, \lambda^-) = \|w(\lambda^+, \lambda^-)\|^2 - 2\alpha \leq \|w(\tau^+, \tau^-)\|^2 - 2$$

2. Since  $(\tau^+, \tau^-)$  satisfy the constraints of problem 1, we have

$$\|w(\sigma^+, \sigma^-)\|^2 \leq \|w(\tau^+, \tau^-)\|^2.$$

## Equivalence of optimization problems continued

3. From (2), we have

$$\|w(\alpha\sigma^+, \alpha\sigma^-)\|^2 \leq \alpha^2 \|w(\tau^+, \tau^-)\|^2 = \|w(\lambda^+, \lambda^-)\|^2$$

4. Subtracting  $2\alpha$  from both sides of this inequality yields

$$\|w(\alpha\sigma^+, \alpha\sigma^-)\|^2 - 2\alpha \leq Q(\lambda^+, \lambda^-).$$

Since  $(\lambda^+, \lambda^-)$  minimize  $Q(\lambda^+, \lambda^-)$ , this inequality must be an equality. Therefore

$$\alpha^2 \|w(\sigma^+, \sigma^-)\|^2 = \alpha^2 \|w(\tau^+, \tau^-)\|^2$$

so  $(\tau^+, \tau^-)$  also gives a minimal value for problem 1.