

Formulating the optimization problem

The optimization problem

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

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

and

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

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

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

The optimization problem (continued)

As above, let our linearly separable sets be

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

$$C(A^\pm) = \left\{ \sum_{i=1}^{n_\pm} \lambda_i^\pm x_i^\pm \mid \right.$$

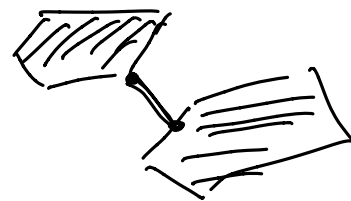
and

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

$$\left. \sum \lambda_i^\pm = 1 \right\} \\ \text{all } \lambda_i \geq 0$$

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

$$\underbrace{w(\lambda^+, \lambda^-)}_{\substack{\text{C}(A^+) \\ \text{C}(A^-)}} = \sum_{i=1}^{n_+} \lambda_i^+ x_i^+ - \sum_{i=1}^{n_-} \lambda_i^- x_i^-.$$



Find λ^\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:

$\|w(\lambda^+, \lambda^-)\|^2$ is a quadratic function in the λ 's with coefficients coming from the dot products of the x_i^\pm .

$$\|w(\lambda^+, \lambda^-)\|^2 = \left(\sum \lambda_i^+ x_i^+ - \sum \lambda_i^- x_i^-, \sum \lambda_i^+ x_i^+ - \sum \lambda_i^- x_i^- \right)$$

typical km $\lambda_i^+ \lambda_j^+ (x_i^+, x_j^+)$

- The constraints are *inequalities* rather than equalities, so a direct application of Lagrange multipliers as we learned in calculus won't work.

$$\sum \lambda_i^+ = 1 \quad \sum \lambda_i^- = 1$$

$$\lambda_i^\pm \geq 0.$$

- 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

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

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

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

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

Equivalence of the reformulated problem

Proof: We have (λ^+, λ^-) solving problem 2 and (σ^+, σ^-) solving problem 1; and finally we have $\underbrace{(\tau^+, \tau^-) = (1/\alpha)(\lambda^+, \lambda^-)}_{\substack{\uparrow \\ Q(\tau^+, \tau^-)}}$. $\|w(\sigma^+, \sigma^-)\|^2 = \sum \sigma_i^{+2} = \sum \sigma_i^{-2}$

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

$$\underbrace{Q(\lambda^+, \lambda^-)}_{\substack{\uparrow \\ Q(\lambda^+, \lambda^-)}} = \underbrace{\|w(\lambda^+, \lambda^-)\|^2}_{\substack{\uparrow \\ Q(\lambda^+, \lambda^-)}} - 2\alpha \leq \underbrace{\|w(\tau^+, \tau^-)\|^2}_{\substack{\uparrow \\ Q(\tau^+, \tau^-)}} - 2 \quad \dots$$

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

$$\underbrace{\|w(\sigma^+, \sigma^-)\|^2}_{\substack{\uparrow \\ Q(\sigma^+, \sigma^-)}} \leq \underbrace{\|w(\tau^+, \tau^-)\|^2}_{\substack{\uparrow \\ Q(\tau^+, \tau^-)}}.$$

Equivalence of optimization problems continued

3. From (2), we have

$$\alpha^2 \|w(\sigma^+, \sigma^-)\|^2 = \underbrace{\|w(\alpha\sigma^+, \alpha\sigma^-)\|^2}_{\leq \alpha^2 \|w(\tau^+, \tau^-)\|^2} = \underbrace{\|w(\lambda^+, \lambda^-)\|^2}_{\star} \quad \alpha\tau^\pm = \lambda^\pm$$

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

$$w(\sigma^+, \sigma^-) = \left(\sum \sigma_i^+ x_i^+ - \sum \sigma_i^- x_i^-, \sum \sigma_i^+ x_i^+ - \sum \sigma_i^- x_i^- \right)$$

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

$$\begin{aligned} \sum \sigma_i^+ &= 1 \quad \sigma_i \geq 0 \\ \sum \sigma_i^- &= 1 \end{aligned}$$

4. Subtracting 2α from both sides of this inequality yields

$$\longrightarrow \underbrace{\|w(\alpha\sigma^+, \alpha\sigma^-)\|^2 - 2\alpha}_{\leq} \leq \underbrace{Q(\lambda^+, \lambda^-)}_{\text{minimum value for } Q}$$

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

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

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

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