Gradient for k-means

Minimization

Exercice 1: Gradient:

Compute the gradient of $J(\Omega, z)$ with respect to Ω and deduce the minimizer Ω^* .

- z is fixed
- ▶ we can see Ω has a vector of \mathbb{R}^{Kd} .

It is sufficient to compute the gradient with respect to all the ω_k separately.

Minimization

We compute the gradient of J with respect to ω_{k_0} .

$$\nabla_{\omega_{k_0}} J = \nabla_{\omega_{k_0}} \sum_{i=1}^{n} \sum_{k=1}^{K} z_i^k ||x_i - \omega_k||^2$$

$$= \sum_{i=1}^{n} \nabla_{\omega_{k_0}} \sum_{k=1}^{K} z_i^k ||x_i - \omega_k||^2$$

$$= \sum_{i=1}^{n} z_i^{k_0} \nabla_{\omega_{k_0}} ||x_i - \omega_{k_0}||^2$$
(1)

Remark: we could compute this also like that:

$$\nabla_{\omega_{k_0}} J = \nabla_{\omega_{k_0}} \sum_{i=1}^n ||x_i - \omega_k||^2$$
 (2)

We need to compute

$$\nabla_{\omega_{k_0}}||x_i - \omega_{k_0}||^2 \tag{3}$$

Let

$$u = \begin{cases} \mathbb{R}^d \to \mathbb{R} \\ y \mapsto ||y||^2 \end{cases}$$
$$v_i = \begin{cases} \mathbb{R}^d \to \mathbb{R} \\ \omega \mapsto x_i - \omega \end{cases}$$
$$w_i = \begin{cases} \mathbb{R}^d \to \mathbb{R} \\ \omega \mapsto ||x_i - \omega||^2 \end{cases}$$

We have

$$w_i = u \circ v_i \tag{4}$$

Differentials

We already know that

$$\forall (y,h) \in (\mathbb{R}^d)^2, du_y(h) = \langle 2y, h \rangle \tag{5}$$

and that

$$\forall (\omega, h) \in (\mathbb{R}^d)^2, d(v_i)_{\omega}(h) = -h \tag{6}$$

By composition,

$$\nabla_{\omega_{k_0}} ||x_i - \omega_{k_0}||^2 = -2(x_i - \omega_{k_0})$$
 (7)

Gradient cancellation

This gradient cancels if

$$\nabla_{\omega_{k_0}} J = \sum_{i=1}^n z_i^{k_0} \nabla_{\omega_{k_0}} ||x_i - \omega_{k_0}||^2$$

$$= -\sum_{i=1}^n z_i^{k_0} 2(x_i - \omega_{k_0})$$

$$= 0$$
(8)

Gradient cancellation

This gradient cancels if

$$2\omega_{k_0} \sum_{i=1}^n z_i^{k_0} = 2\sum_{i=1}^n z_i^{k_0} x_i \tag{9}$$

or equivalently

$$\omega_{k_0} = \frac{\sum_{i=1}^{n} z_i^{k_0} x_i}{\sum_{i=1}^{n} z_i^{k_0}}$$
 (10)

Hence, the minimizer $w_{k_0}^*$ is the average of its cluster.