

The problem

$A = \{a^1, \dots, a^m\} \subset \mathbb{R}^n$ ,  $k \in (1, m)$  is a given fixed number of clusters.

Find  $\{x^1, \dots, x^k\}$  the cluster centers:

$$(C) \min \sum_{i=1}^m \min_{1 \leq l \leq k} d(x^l, a^i) : x^1, \dots, x^k \in \mathbb{R}^n$$

where  $d(\cdot, \cdot)$  is some given distance.

Reformulation & Notations

$$a = (a^1, \dots, a^m) \in (\mathbb{R}^n)^m; \quad a^i \in \mathbb{R}^n \quad i=1, \dots, m$$

$$w = (w^1, \dots, w^k) \in (\mathbb{R}^k)^m; \quad w^i \in \mathbb{R}^k \quad i=1, \dots, m$$

$$X = (x^1, \dots, x^k) \in (\mathbb{R}^n)^k; \quad x^l \in \mathbb{R}^n \quad l=1, \dots, k$$

$$d_i(X) := (d(x^1, a^i), d(x^2, a^i), \dots, d(x^k, a^i)) \in \mathbb{R}^k \quad i=1, \dots, m$$

$$\Delta := \left\{ u \in \mathbb{R}^k : \sum_{l=1}^k u_l = 1, u_l \geq 0 \quad l=1, \dots, k \right\}$$

$$\langle u, v \rangle = \sum_{l=1}^k u_l v_l \quad \forall u, v \in \mathbb{R}^k$$

$$\text{using the fact that: } \min_{1 \leq l \leq k} u_l = \min \{ \langle u, v \rangle : v \in \Delta \}$$

$\Rightarrow$  Smooth Reformulation of (C):

$$(C) \min_{X \in (\mathbb{R}^n)^k} \sum_{i=1}^m \min \{ \langle w^i, d^i(X) \rangle : w^i \in \Delta \}$$

$$\Leftrightarrow (C) \min_{\substack{X, w \\ (\mathbb{R}^n)^k \quad (\mathbb{R}^k)^m}} \sum_{i=1}^m \left\{ \langle w^i, d^i(X) \rangle + f_{\Delta}(w^i) \right\}$$

①

Define for each  $i=1, \dots, m$ :

$$H_i(w, x) = \langle w^i, d^i(x) \rangle$$

$$G(w^i) = f_{\Delta}(w^i)$$

and

$$H(w, x) = \sum_{i=1}^m H_i(w, x) ; \quad G(w) = \sum_{i=1}^m G(w^i)$$

Then (C)  $\Leftrightarrow \min \{ G(w) + H(w, x) : w \in \mathbb{R}^{km}, x \in \mathbb{R}^{nk} \}$

Setting  $d(u, v) := \|u - v\|^2$

PALM for (C):

0. Initialization:  $t=0$ , and pick  $(w(0), x(0))$

1. Cluster Assignment: For  $i=1, \dots, m$  solve (parallelly):

$$\begin{aligned} w^i(t+1) &= \operatorname{argmin}_{w^i \in \Delta} \left\{ \langle w^i, d^i(x(t)) \rangle + \frac{\alpha(t)}{2} \|w^i - w^i(t)\|^2 \right\} = \\ &= \Pi_{\Delta} \left( w^i(t) - \frac{d^i(x(t))}{\alpha(t)} \right) \end{aligned}$$

Here  $\alpha(t) > 0$  is the stepsize that is chosen to be

$$\alpha(t) = \min_{1 \leq l \leq k} \sum_{i=1}^m w_l^i$$

2. Cluster Centers: Compute  $x^l \in \mathbb{R}^n$   $l=1, \dots, k$  via

$$x(t+1) = \operatorname{argmin} \{ H(w(t+1), x) : x \in \mathbb{R}^{nk} \} \Rightarrow$$

$$x^l(t+1) = \frac{\sum_{i=1}^m w_l^i(t+1) \cdot a^i}{\sum_{i=1}^m w_l^i(t+1)}$$

Setting  $\Psi(w, x) := H(w, x) + G(w)$  and  $z := (w, x)$

You should remind that  $G(w(t)) = 0$  for any  $t > 0$  and thus

$$\Psi(z^*(t)) = H(z^*(t)) \quad \text{for all } t \geq 0.$$

Sufficient decrease property

There is missing power 2 here

$$\exists \rho_1 > 0 : \|z(t+1) - z(t)\|^2 \leq \Psi(z(t)) - \Psi(z(t+1))$$

proof: from step 1:

$$\langle w^i(t+1), d^i(x(t)) \rangle + \frac{\alpha(t)}{2} \|w^i(t+1) - w^i(t)\|^2 \leq$$

$$\langle w^i(t), d^i(x(t)) \rangle + \frac{\alpha(t)}{2} \|w^i(t) - w^i(t)\|^2$$

0

$$\Rightarrow \frac{\alpha(t)}{2} \|w^i(t+1) - w^i(t)\|^2 \leq H_i(w(t), x(t)) - H_i(w(t+1), x(t))$$

$$\Rightarrow \frac{\alpha(t)}{2} \|w(t+1) - w(t)\|^2 = \frac{\alpha(t)}{2} \sum_{i=1}^m \|w^i(t+1) - w^i(t)\|^2 \leq$$

$$\leq \sum_{i=1}^m H_i(w(t), x(t)) - \sum_{i=1}^m H_i(w(t+1), x(t)) = H(w(t), x(t)) - H(w(t+1), x(t))$$

just why there exists  $m > 2$ . The function  $H(w, x)$  is explicitly given so you can compute the parameter

~~$H(w, x)$  strongly convex~~

Since  $x \mapsto H(w, x)$  is strongly convex then  $\exists m > 0$ :

$$H(w(t+1), x(t)) - H(w(t+1), x(t+1)) \geq$$

$$\geq \nabla_x H(w(t+1), x(t+1))^T (x(t) - x(t+1)) + \frac{m}{2} \|x(t) - x(t+1)\|^2 =$$

$x(t+1)$  is  $\rightarrow$  argmin of  $\{H(w(t+1), x) \mid x \in (\mathbb{R}^n)^K\}$

$$= \frac{m}{2} \|x(t) - x(t+1)\|^2$$

Taking  $\frac{\alpha(t)}{2} = \rho_1 = \frac{m}{2}$  then:

$$\rho_1 \|z(t+1) - z(t)\|^2 \leq \rho_1 (\|w(t+1) - w(t)\|^2 + \|x(t+1) - x(t)\|^2) \leq$$

$$\leq [H(w(t), x(t)) - H(w(t+1), x(t))] + [H(w(t+1), x(t)) - H(w(t+1), x(t+1))]$$

$$= H(z(t)) - H(z(t+1)) = \Psi(z(t)) - \Psi(z(t+1))$$

$w(t), w(t+1) \in \Delta$

This is not true, you should define

$$\rho_i = \min \left\{ \frac{\alpha(t)}{2}, \frac{m}{2} \right\}$$

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Lemma: At step  $t$ ,  $x \rightarrow H(w(t), x)$  is strongly convex  
if and only if  $\min_{1 \leq l \leq k} \sum_{i=1}^m w_l^i > 0$ .

proof: Since  $x \rightarrow H(w(t), x)$  is  $C^2$ , it is strongly convex iff its Hessian matrix smallest e.v. is positive.

$$\nabla_{x_s} \nabla_{x_l} H(w(t), x) = \begin{cases} 0 & s \neq l \\ 2 \sum_{i=1}^m w_l^i(t) & s = l \end{cases}$$

$\Rightarrow x \rightarrow H(w(t), x)$  is strongly convex iff  $w_l = \min_{1 \leq l \leq k} \sum_{i=1}^m w_l^i > 0$

You may write all the needed properties of  $H$  before you start your proofs and thus you will use them in the sequel.

Remark: if at step  $t$   $\min_{1 \leq l \leq k} \sum_{i=1}^m w_l^i = 0$  then we can conclude that the number of clusters can be reduced by 1. Hence w.l.o.g. we assume that at each step  $t$   $\min_{1 \leq l \leq k} \sum_{i=1}^m w_l^i > 0$

You might give more details here, like that in this case  $w_l^i = 0$  for all  $i$ .

Subgradient lower bound for iterates property:

$\exists \rho_z > 0 : \|x(t+1)\| \leq \rho_z \|z(t+1) - z(t)\|$   
where  $f(t) \in \partial \Psi(z(t))$ ,  $z := (w, x)$

proof:  $\Psi = H + G$

*You don't need to write  $z$  here*

$$\partial \Psi = \nabla_x H + \partial_x G = (\nabla_w H, \nabla_x H) + (\partial_{w^1} f_\Delta, \dots, \partial_{w^m} f_\Delta, 0, \dots, 0)$$

$$= ((\nabla_{w^i} H + \partial_{w^i} f_\Delta)_{i=1, \dots, m}, \nabla_x H)$$

*also here*

$$\Rightarrow \partial \Psi(z(t+1)) = ((\nabla_{w^i} H(w(t+1), x(t+1)) + \partial_{w^i} f_\Delta(w^i(t+1)))_{i=1, \dots, m}, \nabla_x H(w(t+1), x(t+1)))$$

$\parallel$   
This is true but you should give reference for that

(4)

You might add here another equality saying that since  $\nabla_{w^i} H_i(z(t+1)) = d_i^i(x(t+1))$  we get that  $\partial \Psi(z(t+1)) = (d^i(x(t+1)) + \partial_{w^i} f_\Delta(w^i(t+1)))_{i=1, \dots, m}, 0)$

according to my previous remark  $\rightarrow \partial^i(x(t+1))$

$$\Rightarrow \|\partial\psi(z(t+1))\| \leq \sum_{i=1}^m \|\nabla_{w_i} H(w(t+1), x(t+1)) + \partial_{w_i} f_{\Delta}(w_i(t+1))\|$$

(i) Now, since  $w(t+1) = \underset{w \in \mathbb{R}^k}{\operatorname{argmin}} \left\{ \langle w^i, d^i(x(t)) \rangle + \frac{\alpha(t)}{2} \|w^i - w^i(t)\|^2 + f_{\Delta}(w^i) \right\}$

$$\Rightarrow \exists w^i(t+1) \in \partial f_{\Delta}(w^i(t+1)) : d^i(x(t)) + \alpha(t)(w^i(t+1) - w^i(t)) + w^i(t+1) = 0$$

$$\Rightarrow \|\partial\psi(z(t+1))\| \leq \sum_{i=1}^m \|d^i(x(t+1)) - d^i(x(t)) - \alpha(t)(w^i(t+1) - w^i(t))\| \leq$$

$$\leq \sum_{i=1}^m \|d^i(x(t+1)) - d^i(x(t))\| + m\alpha(t) \|z(t+1) - z(t)\|$$

You can bound it by a better constant  $\sqrt{m}$  instead of  $m$ . Show it!

Hence, it is sufficient to show that  $d^i(\cdot)$  is Lipschitz Continuous

i.e.  $\exists M \geq 0 : \|d^i(x) - d^i(y)\| \leq M \|x - y\| \quad x, y \in (\mathbb{R}^n)^k$

Proposition: if  $\|\nabla f\|$  is bounded on  $\mathcal{N}$ ,  $M := \sup_{z \in \mathcal{N}} \|\nabla f(z)\|$

then  $\|f(x) - f(y)\| \leq M \|x - y\| \quad \forall x, y \in \mathcal{N}$

Lemma:  $\|\nabla_x d^i(\cdot)\|$  is bounded on  $\mathcal{N} = \{x \in (\mathbb{R}^n)^k \mid \|x^i\| \leq \sum_{j=1}^m \|a^j\| \}$

proof:  $\|\nabla_x d^i(x)\| = \|(x^1 - a^1), \dots, (x^k - a^k)\| \leq 2 \sum_{l=1}^k \|x^l - a^l\| \leq$   
 $\leq 2 \left( \sum_{l=1}^k \|x^l\| + k \|a^i\| \right) \leq 2 \left( k \sum_{j=1}^m \|a^j\| + k \|a^i\| \right) \leq 2(k+1) \sum_{j=1}^m \|a^j\|$   
 $\quad \quad \quad \uparrow$   
 $\quad \quad \quad x \in \mathcal{N}$

Lemma:  $x(t) \in \mathcal{N} \quad \forall t \geq 1$

proof:  $\|x^i(t+1)\| = \left\| \sum_{j=1}^m \frac{w_j^i(t+1) a^j}{\sum_{j=1}^m w_j^i(t+1)} \right\| \leq \sum_{j=1}^m \frac{w_j^i(t+1)}{\sum_{j=1}^m w_j^i(t+1)} \|a^j\| \leq \sum_{j=1}^m \|a^j\|$

$\Rightarrow$  It follows that  $\rho_2 = m(M + \alpha(t))$  is an appropriate constant, i.e.  $\|x(t+1)\| \leq \rho_2 \|z(t+1) - z(t)\|, \delta(t) \in \partial\psi(z(t))$

You can improve it according to what I have mentioned above!

You don't need result as is because anyway you have to prove this result another the assumption that ~~the~~ the generated sequence is bounded and in this case it is clear that (5)

$$\|\nabla_x d^i(x(t))\| \text{ is bounded}$$

for all  $t \geq 0$ .

~~we can improve~~

