

# 1 Online Sinkhorn

The Sinkhorn objective rewrites

$$\max_{f, g \in \mathcal{C}(\mathcal{X})} \langle f, \alpha \rangle + \langle g, \beta \rangle - \varepsilon \langle \alpha \otimes \beta, \exp(-\frac{f \oplus g - C}{\varepsilon}) \rangle$$

We perform the following change of variable  $\mu = \alpha e^{f/\varepsilon}$ ,  $\nu = \beta e^{g/\varepsilon}$ , to obtain the equivalent problem, in  $\mathcal{M}^+(\mathcal{X})$

$$\min_{\mu, \nu \in \mathcal{M}^+(\mathcal{X})} \text{KL}(\alpha|\mu) + \text{KL}(\beta|\nu) + \varepsilon \langle \mu \otimes \nu, \exp(-\frac{C}{\varepsilon}) \rangle \triangleq f(\mu, \nu).$$

The problem is not jointly convex, but convex in  $\mu$  and  $\nu$ . We may approach this problem from a game point of view of finding a local Nash equilibrium  $(\mu^*, \nu^*)$  such that

$$\begin{aligned} \mu^* &= \underset{\mu \in V(\mu^*)}{\operatorname{argmin}} \mathcal{F}(\mu, \nu^*) \\ \nu^* &= \underset{\nu \in V(\nu^*)}{\operatorname{argmin}} \mathcal{F}(\mu^*, \nu), \end{aligned} \tag{1}$$

where  $V$  are open sets. Such a formalism is useful as results on mirror descent convergence in multi-agent setting exist for this problem. To solve (1), we need to define distance generating functions to move back and forth from  $\mu$  and  $\nu$  and their dual form. We define

$$\begin{aligned} \omega_\alpha(\mu) &\triangleq \text{KL}(\alpha|\mu) \\ \omega_\beta(\nu) &\triangleq \text{KL}(\beta|\nu) \end{aligned}$$

, associated the the mirror maps

$$\begin{aligned} \nabla_\mu \omega_\alpha(\mu) &= -\frac{d\alpha}{d\mu} \quad (= -\exp(-f/\varepsilon)), \\ \nabla_\nu \omega_\beta(\nu) &= -\frac{d\beta}{d\nu} \quad (= -\exp(-g/\varepsilon)), \end{aligned}$$

with inverse

$$\begin{aligned} \nabla_\mu \omega_\alpha^*(p) &= -\frac{\alpha}{p}, \\ \nabla_\nu \omega_\beta^*(q) &= -\frac{\beta}{q}. \end{aligned}$$

**Algorithm.** Let us consider the simple simultaneous mirror descent setting, where we build the sequence of iterate  $(\mu_t, \nu_t)_t$ . It is easy to shows that if we start from  $\mu_0 \gg \alpha$  and  $\nu_0 \gg \beta$ , the iterates will remain absolutely continuous with respect to  $\alpha$  and  $\beta$ . We will therefore write  $\mu_t = \alpha e^{f_t/\varepsilon}$ ,  $\nu_t = \beta e^{g_t/\varepsilon}$ . The mirror descent iterations rewrite (for  $\mu$ )

$$\mu_{t+1} = \frac{\alpha}{e^{-f_t/\varepsilon} + \eta \nabla_\mu \mathcal{F}(\mu_t, \nu_t)},$$

with  $\nabla_\mu \mathcal{F}(\mu_t, \nu_t) = -\exp(-\frac{f_t}{\varepsilon}) + \varepsilon \int_y \exp(\frac{g(y) - C(\cdot, y)}{\varepsilon}) d\beta(y)$ . We therefore have the following update rules

$$\begin{aligned} \exp(-\frac{f_{t+1}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{f_t}{\varepsilon}) + \eta \mathbb{E}_\beta[\varepsilon \exp(\frac{g_t(y) - C(\cdot, y)}{\varepsilon})], \\ \exp(-\frac{g_{t+1}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{g_t}{\varepsilon}) + \eta \mathbb{E}_\alpha[\varepsilon \exp(\frac{f_t(x) - C(x, \cdot)}{\varepsilon})]. \end{aligned}$$

Assuming we sample  $\hat{\beta}_t = \sum_{i=1}^n b_{i,t} \delta_{y_{i,t}}$  and  $\hat{\alpha}_t = \sum_{i=1}^n a_{i,t} \delta_{x_{i,t}}$ , we can approximate the expectations above, and expect, with decreasing step-sizes to achieve convergence.

Some variants (more likely to converge better) may be considered. The alternated variant writes

$$\begin{aligned}\exp(-\frac{f_{t+1}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{f_t}{\varepsilon}) + \eta \mathbb{E}_\beta[\varepsilon \exp(\frac{g_t(y) - C(\cdot, y)}{\varepsilon})], \\ \exp(-\frac{g_{t+1}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{g_t}{\varepsilon}) + \eta \mathbb{E}_\alpha[\varepsilon \exp(\frac{f_{t+1}(x) - C(x, \cdot)}{\varepsilon})],\end{aligned}$$

and the extrapolated version

$$\begin{aligned}\exp(-\frac{f_{t+1/2}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{f_t}{\varepsilon}) + \eta \mathbb{E}_\beta[\varepsilon \exp(\frac{g_t(y) - C(\cdot, y)}{\varepsilon})], \\ \exp(-\frac{g_{t+1/2}}{\varepsilon}) &= (1 - \eta) \exp(-\frac{g_t}{\varepsilon}) + \eta \mathbb{E}_\alpha[\varepsilon \exp(\frac{f_t(x) - C(x, \cdot)}{\varepsilon})], \\ \exp(-\frac{f_{t+1}}{\varepsilon}) &= \exp(-\frac{f_t}{\varepsilon}) - \eta \exp(-\frac{g_{t+1/2}}{\varepsilon}) + \eta \mathbb{E}_\beta[\varepsilon \exp(\frac{g_{t+1/2}(y) - C(\cdot, y)}{\varepsilon})], \\ \exp(-\frac{g_{t+1}}{\varepsilon}) &= \exp(-\frac{g_t}{\varepsilon}) - \eta \exp(-\frac{f_{t+1/2}}{\varepsilon}) + \eta \mathbb{E}_\alpha[\varepsilon \exp(\frac{f_{t+1/2}(x) - C(x, \cdot)}{\varepsilon})].\end{aligned}$$

**Computations.** In the simple simultaneous case, we can track  $f_t$  in memory by the following representation

$$\begin{aligned}f_t(\cdot) &= -\varepsilon \log \sum_{s=0}^t w_{t,s} \sum_{j=1}^n b_{s,j} \exp\left(g_{s-1}(y_{s,j}) - \frac{C(\cdot, y_{s,j})}{\varepsilon}\right) \\ g_t(\cdot) &= -\varepsilon \log \sum_{s=0}^t w_{t,s} \sum_{i=1}^n a_{s,i} \exp\left(f_{s-1}(x_{s,i}) - \frac{C(x_{s,i}, \cdot)}{\varepsilon}\right),\end{aligned}$$

with  $w_{t,s} = \eta(1 - \eta)^{t-s}$  for  $1 \leq s \leq t$ ,  $w_{t,0} = (1 - \eta)^t$ , and we set  $g_{-1} = f_{-1} = 0$ . The weights are a bit more complex as  $\eta$  depends on time.

The alternated version sets

$$g_t(\cdot) = -\varepsilon \log \sum_{s=0}^t w_{t,s} \sum_{i=1}^n a_{s,i} \exp\left(f_s(x_{s,i}) - \frac{C(x_{s,i}, \cdot)}{\varepsilon}\right),$$

Setting  $q_{t,s,i} = w_{t,s} b_{s,j} \exp(g_{s-1}(y_{s,j}))$  and  $p_{t,s,i} = w_{t,s} b_{s,j} \exp(f_{s-1}(x_{s,j})/\varepsilon)$ , we can derive simple update rules for  $p$  and  $q$ :

$$p_{t,t,i} = \eta b_{t,j} \exp(f_{t-1}(x_{t,j})), \quad \forall s < t, \quad p_{t,s,i} = (1 - \eta) p_{t-1,s,i}$$

## 2 Analysis

Bregman divergence associated to  $\varphi$  (désolé j'ai changé de notation).

$$d_\varphi(f_2|f_1) = \langle \alpha, \exp(\frac{f_2 - f_1}{\varepsilon}) - 1 - \frac{f_2 - f_1}{\varepsilon} \rangle \geq \langle \alpha, (\frac{f_2 - f_1}{2\varepsilon})^2 \rangle$$

For convergence of MD on  $\min f(x)$  with mirror map  $\varphi$ , we need to show, according to Gabriel, Jalal and Kelvin

$$\mu d_\varphi(x_2|x_1) \leq d_f(x_2|x_1) \leq L d_\varphi(x_2|x_1).$$

Can we use that here ? Beware that we are in an alternated setting

### 2.1 Sketch of proof

See proof of Th2 in Ya Ping's paper.

### 2.1.1 General proof

By using the dual iteration and the three point property (normally holds by def of  $D_\alpha$  and  $D_\beta$ ):

$$\begin{aligned}\langle \mu_t - \mu, -\nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle &= \frac{1}{\eta} \langle \mu_t - \mu, \nabla_\mu w_\alpha(\mu_{t+1}) - \nabla_\mu w_\alpha(\mu_t) \rangle \\ &= \frac{1}{\eta} [D_{w_\alpha}(\mu, \mu_t) - D_{w_\alpha}(\mu, \mu_{t+1}) + D_{w_\alpha}(\mu_t, \mu_{t+1})]\end{aligned}$$

Suppose we can show (TO DO):

$$D_{w_\alpha}(\mu_t, \mu_{t+1}) \leq \eta^2 M^2$$

Then we have:

$$\begin{aligned}\frac{1}{T} \sum_{t=1}^T \langle \mu_t - \mu, -\nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle &= \sum_{t=1}^T \frac{1}{\eta} [D_{w_\alpha}(\mu, \mu_t) - D_{w_\alpha}(\mu, \mu_{t+1}) + D_{w_\alpha}(\mu_t, \mu_{t+1})] \\ &\leq \frac{D_{w_\alpha}(\mu, \mu_1)}{\eta} + \eta M^2 T\end{aligned}$$

Similarly:

$$\frac{1}{T} \sum_{t=1}^T \langle \nu_t - \nu, -\nabla_\nu \mathcal{F}(\mu_t, \nu_t) \rangle \leq \frac{D_{w_\beta}(\mu, \mu_1)}{\eta} + \eta M^2 T$$

Summing up the two previous equations and replacing  $(\mu, \nu)$  by  $(\mu^*, \nu^*)$ , we get:

$$\frac{1}{T} \sum_{t=1}^T \langle \mu_t - \mu^*, -\nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle + \langle \nu_t - \nu^*, -\nabla_\nu \mathcal{F}(\mu_t, \nu_t) \rangle \leq \frac{D_0}{\eta} + 2\eta M^2 T$$

where  $D_0 = D_{w_\alpha}(\mu^*, \mu_1) + D_{w_\beta}(\nu^*, \nu_1)$ .

Then, by optimality of  $\mu^*$  and convexity of  $\mathcal{F}$ :

$$\mathcal{F}(\mu_t, \nu_t) - \mathcal{F}(\mu^*, \nu^*) \leq \mathcal{F}(\mu_t, \nu_t) - \mathcal{F}(\mu^*, \nu_t) \leq \langle \mu^* - \mu_t, \nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle = \langle \mu_t - \mu^*, -\nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle$$

Hence:

$$\frac{1}{T} \sum_{t=1}^T \mathcal{F}(\mu_t, \nu_t) - \mathcal{F}(\mu^*, \nu^*) \leq \frac{D_0}{\eta} + 2\eta^2 M^2 T$$

### 2.1.2 Tricky part

Now let's try to prove Equation 2.1.1. What we need is

- $D_{w_\alpha}(\mu, \nu) = D_{w_\alpha^*}(\nabla_\mu w_\alpha(\mu), \nabla_\mu w_\alpha(\nu))$  (eq A.13 in Ya Ping's paper)
- *relative smoothness of  $w_\alpha^*$  wrt  $\|\cdot\|_\infty$*  (eq A.11 and A.12 in Ya Ping's paper)

If it's true:

$$\begin{aligned}D_{w_\alpha}(\mu_t, \mu_{t+1}) &= D_{w_\alpha^*}(\nabla_\mu w_\alpha(\mu_t), \nabla_\mu w_\alpha(\mu_{t+1})) \leq \|\nabla_\mu w_\alpha(\mu) - \nabla_\mu w_\alpha(\nu)\|_\infty^2 \\ &= \|\exp(-\frac{f_{t+1}}{\varepsilon}) - \exp(-\frac{f_t}{\varepsilon})\|_\infty^2 = \eta^2 \|\nabla \mathcal{F}_\mu(\mu_t, \nu_t)\|_\infty^2\end{aligned}$$

and

$$\|\nabla \mathcal{F}_\mu(\mu_t, \nu_t)\|_\infty^2 \leq ?$$

WIP

$$D_{w_\alpha}(\mu_t, \mu_{t+1}) \geq \|\log \frac{d\mu_{t+1}}{d\mu_t}\|_\alpha^2$$

We can show

$$\begin{aligned} D_{\omega_\alpha}(\mu_t | \mu_{t+1}) &= \eta^2 \langle \alpha, 1 - \exp(\frac{f_t - \hat{f}_{t+1}}{\varepsilon}) \rangle = \eta^2 (1 - \langle \alpha, \nabla_\mu \mathcal{F}(\mu_t, \nu_t) \rangle) \\ \hat{f}_{t+1}(\cdot) &= -\varepsilon \log \int_y \exp(\frac{g_t(y) - C(\cdot, y)}{\varepsilon}) d\beta(y). \end{aligned}$$

Avec Sinkhorn sans bruit  $f_t - \hat{f}_{t+1}$  va rester tranquille.

### 3 Proof of convergence

We want to solve

$$\min_{\mu \in \mathcal{M}^+(\mathcal{X}), \nu \in \mathcal{M}^+(\mathcal{X})} F(\mu, \nu) \triangleq \text{KL}(\alpha | \mu) + \text{KL}(\beta | \nu) + \langle \mu \otimes \nu, \exp(-C) \rangle$$

Let's write  $x = (\mu, \nu)$  and  $F(\mu, \nu) = F(x)$  the objective. We define the iterates  $x_t = (\mu_t, \nu_t)$ ,  $x_{t+1/2} = (\mu_{t+1}, \nu_t)$ ,  $x_t = (\mu_{t+1}, \nu_{t+1})$ . First note that we have

$$D_{F(\mu, \cdot)} = D_{\omega_\alpha(\cdot)} \quad D_{F(\cdot, \nu)} = D_{w_\beta(\cdot)},$$

so that at every iteration, we perform a mirror step with a function that is both 1-relatively smooth and 1-relatively strongly convex.

Let  $\nu$  be fixed, and let us define  $F_\nu(\cdot) = F(\cdot, \nu)$ . From the smoothness of  $F_\nu(\cdot)$  and from its convexity we have, for all  $\mu_x, \mu_y, \mu_z \gg \alpha$ ,

$$\begin{aligned} F(\mu_x, \nu) &\leq F(\mu_y, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_x - \mu_y \rangle + LD_{\omega_\alpha}(\mu_x, \mu_y), \\ F(\mu_y, \nu) &\leq F(\mu_z, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_y - \mu_z \rangle \end{aligned}$$

Combining both, we obtain

$$\begin{aligned} F(\mu_x, \nu) &\leq F(\mu_z, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_x - \mu_z \rangle + LD_{\omega_\alpha}(\mu_x, \mu_y) \\ \langle \nabla F(\mu_y, \nu), \mu_x - \mu_z \rangle &\geq F(\mu_x, \nu) - F(\mu_z, \nu) - LD_{\omega_\alpha}(\mu_x, \mu_y). \end{aligned}$$

We now use the three point property:

$$D_{\omega_\alpha}(\mu_z, \mu_y) - D_{\omega_\alpha}(\mu_z, \mu_x) - D_{\omega_\alpha}(\mu_x, \mu_y) = \langle \nabla \omega_\alpha(\mu_x) - \nabla \omega_\alpha(\mu_y), \mu_z - \mu_x \rangle,$$

replacing  $\mu_y = \mu_k, \mu_x = \mu_{k+1}, \nu = \nu_k$ , we obtain from the definition of the gradient update

$$\begin{aligned} D_{\omega_\alpha}(\mu_z, \mu_k) - D_{\omega_\alpha}(\mu_z, \mu_{k+1}) - D_{\omega_\alpha}(\mu_{k+1}, \mu_k) &= \eta_k \langle \nabla_\mu F(\mu_k, \nu_k), \mu_{k+1} - \mu_z \rangle + \eta_k \langle \xi_k, \mu_{k+1} - \mu_z \rangle \\ &\geq \eta_k (F(\mu_{k+1}, \nu_k) - F(\mu_z, \nu_k)) - L\eta_k D_{\omega_\alpha}(\mu_{k+1}, \mu_k) \\ &\quad + \eta_k \langle \xi_k, \mu_{k+1} - \mu_z \rangle. \end{aligned}$$

Hence, mimicking the derivation for  $\nu$

$$\begin{aligned} \eta_k (F(\mu_{k+1}, \nu_{k+1}) - F(\mu_{k+1}, \nu_k)) &\leq D_{\omega_\beta}(\nu_z, \nu_k) - D_{\omega_\beta}(\nu_z, \nu_{k+1}) - (1 - \eta_k L) D_{\omega_\beta}(\nu_{k+1}, \nu_k). \\ \eta_k (F(\mu_{k+1}, \nu_k) - F(\mu_z, \nu_k)) &\leq D_{\omega_\alpha}(\mu_z, \mu_k) - D_{\omega_\alpha}(\mu_z, \mu_{k+1}) - (1 - \eta_k L) D_{\omega_\alpha}(\mu_{k+1}, \mu_k) \end{aligned}$$

Setting  $\mu_z = \mu_k, \nu_z = \nu_k$ , we obtain a descent lemma.

$$F(\mu_{k+1}, \nu_{k+1}) \leq F(\mu_{k+1}, \nu_k) \leq F(\mu_k, \nu_k),$$

which ensures almost sure convergence of  $F(\mu_k, \nu_k)$  to  $F^*$  using constant step-sizes (gradient is zero for  $k \rightarrow \infty$ ).

We further have, replacing  $\mu_z = \mu^*, \nu_z = \nu^*$

$$F(\mu_{k+1}, \nu_{k+1}) - F^* - \frac{\sum_{k=1}^k \eta_k (F(\mu_{k+1}, \nu^*) + F(\mu^*, \nu_k) - 2F^*)}{2 \sum_{k=1}^k \eta_k} \leq \frac{D_{\omega_\alpha}(\mu^*, \mu_1) + D_{\omega_\beta}(\nu^*, \nu_1)}{2 \sum_{k=1}^K \eta_k}$$

Now note that

$$(F(\mu_{k+1}, \nu^*) + F(\mu^*, \nu_k) - 2F^*) = D_{\omega_\alpha}(T(g_k, \beta), f^*) + D_{\omega_\beta}(T(f_{k+1}, \alpha), g^*)$$

We may show the contractance of the soft c-tranform for the following metric

$$\varphi(f, g) = \min_{f^*, g^* \in \mathcal{S}} D_{\omega_\alpha}(f, f^*) + D_{\omega_\beta}(g, g^*).$$

Namely, if

$$\varphi(T(f, \alpha), T(g, \beta)) \leq \varphi(f, g).$$

It is then easy to show (convexity argument) that

$$\varphi(f_{t+1}, g_{t+1}) \leq (1 - \eta) \varphi(f_t, g_t) + \eta \varphi(T(f_t, \alpha), T(g_t, \beta))$$

What can be shown is unfortunately

$$D_{\omega_\alpha}(T(g, \beta), f^*) + D_{\omega_\beta}(T(f, \alpha), g^*) \leq D_{\omega_\alpha}(f^*, f) + D_{\omega_\beta}(g^*, g).$$

**Simultaneous gradient descent.** We have

$$\begin{aligned} \eta_k (F(\mu_k, \nu_{k+1}) - F(\mu_k, \nu_z)) &\leq D_{\omega_\beta}(\nu_z, \nu_k) - D_{\omega_\beta}(\nu_z, \nu_{k+1}) - (1 - \eta_k L) D_{\omega_\beta}(\nu_{k+1}, \nu_k). \\ \eta_k (F(\mu_{k+1}, \nu_k) - F(\mu_z, \nu_k)) &\leq D_{\omega_\alpha}(\mu_z, \mu_k) - D_{\omega_\alpha}(\mu_z, \mu_{k+1}) - (1 - \eta_k L) D_{\omega_\alpha}(\mu_{k+1}, \mu_k) \end{aligned}$$

Therefore

$$\begin{aligned} \eta_k (F(\mu_k, \nu_{k+1}) + F(\mu_{k+1}, \nu_k) - 2F(\mu_z, \nu_z)) \\ - (F(\mu_k, \nu_z) + F(\mu_z, \nu_k) - 2F(\mu_z, \nu_z)) &\leq D_{\omega_\beta}(\nu_z, \nu_k) + D_{\omega_\alpha}(\mu_z, \mu_k) \\ &\quad - (D_{\omega_\alpha}(\mu_z, \mu_{k+1}) + D_{\omega_\beta}(\nu_z, \nu_{k+1})) \\ &\quad - (1 - \eta_k) (D_{\omega_\beta}(\nu_{k+1}, \nu_k) + D_{\omega_\alpha}(\mu_{k+1}, \mu_k)) \end{aligned}$$

Let's observe that, for all  $(\mu^*, \nu^*) \in \mathcal{S}$ ,  $(\mu, \nu)$

$$\begin{aligned} D_{\omega_\beta}(\nu^*, \nu) + D_{\omega_\alpha}(\mu^*, \mu) &= \text{KL}(\alpha|\mu) + \text{KL}(\beta|\nu) + (\langle \mu \otimes \nu^*, \exp(-C) \rangle + \langle \mu^* \otimes \nu, \exp(-C) \rangle - 2) - F(\mu^*, \nu^*) \\ &= F(\mu, \nu^*) + F(\mu^*, \nu) - 2F(\mu^*, \nu^*) \end{aligned}$$

We take  $(\mu_z, \nu_z) = (\mu^*, \nu^*)$ , that optimizes

$$\begin{aligned} G(\mu_k, \nu_k) &\triangleq \min_{\mu^*, \nu^*} D_{\omega_\beta}(\nu^*, \nu_k) + D_{\omega_\alpha}(\mu^*, \mu_k) \\ &= \text{KL}(\alpha|\mu) + \text{KL}(\beta|\nu) + 2(\sqrt{\langle \mu^* \otimes \nu, \exp(-C) \rangle \langle \mu \otimes \nu^*, \exp(-C) \rangle} - 1) - F^* \end{aligned}$$

Then

$$\begin{aligned} \eta_k (F(\mu_k, \nu_{k+1}) + F(\mu_{k+1}, \nu_k) - 2F^*) &\leq (1 + \eta_k) G(\mu_k, \nu_k) - G(\mu_{k+1}, \nu_{k+1}) \\ &\quad - (1 - \eta_k) (D_{\omega_\beta}(\nu_{k+1}, \nu_k) + D_{\omega_\alpha}(\mu_{k+1}, \mu_k)) \end{aligned}$$

Note that, using  $\sqrt{ab} \geq \frac{a+b}{2}$

$$F(\mu_k, \nu_{k+1}) + F(\mu_{k+1}, \nu_k) - G(\mu_k, \nu_k) - 2F^* \geq \text{KL}(\alpha|\mu_{k+1}) + \text{KL}(\beta|\nu_{k+1}) - F^* + 2u_k,$$

where

$$u_k \triangleq \frac{\langle \mu_k \otimes (\nu_{k+1} - \nu^*), \exp(-C) \rangle + \langle (\mu_{k+1} - \mu^*) \otimes \nu_k, \exp(-C) \rangle}{2}$$

Now let's observe that the harmonic mean is always smaller than the arithmetic mean:

$$\frac{d\mu_{k+1}}{d\alpha} = \frac{1}{(1-\eta_k) \frac{1}{\frac{d\mu_k}{d\alpha}} + \eta_k \frac{1}{\frac{d\alpha \exp(T(g_k, \beta))}{d\alpha}}} \leq (1-\eta_k) \frac{d\mu_k}{d\alpha} + \eta_k \frac{d\alpha \exp(T(g_k, \beta))}{d\alpha},$$

hence

$$\begin{aligned} \mu_{k+1} &\leq (1-\eta_k)\mu_k + \eta_k T(\nu_k, \beta) \\ \nu_{k+1} &\leq (1-\eta_k)\nu_k + \eta_k T(\mu_k, \beta) \end{aligned}$$

Therefore, from the definition of the  $c$ -transform

$$u_{k+1} \leq (1-\eta_k)u_k$$

Finally

$$\eta_k (\text{KL}(\alpha|\mu_{k+1}) + \text{KL}(\beta|\nu_{k+1}) - F^*) \leq G(\mu_k, \nu_k) - G(\mu_{k+1}, \nu_{k+1}) - 2\eta_k \mu_k,$$

hence convergence !

## 4 Proofs

### 4.1 A simple "local" convergence proof in the non-noisy case

Remark that

$$\begin{aligned} f_{k+1} &= -\log(\exp(-f_k)(1-\eta_k) + \eta_k \exp(-T(g_k, \beta))) \\ g_{k+1} &= -\log(\exp(-g_k)(1-\eta_k) + \eta_k \exp(-T(f_{k+1}, \alpha))) \end{aligned}$$

Let  $(f^*, g^*)$  be a coupl of solution. There exists  $x \in \mathcal{X}$  such that  $\|f_{k+1} - f^*\|_{\text{var}} = |f_{k+1}(x) - f^*(x)|$ . For this  $x$ , using the convexity of  $-\log(x)$  (more or less the mirror map),

$$\begin{aligned} \|f_{k+1} - f^*\|_{\text{var}} &= |-\log((1-\eta_k) \exp(f^* - f_k(x)) + \eta_k \exp(f^* - T(g_k, \beta)(x)))| \\ &\leq (1-\eta_k) |f_k(x) - f^*(x)| + \eta_k |T(g_k, \beta)(x) - f^*(x)| \\ &\leq (1-\eta_k) \|f_k - f^*\|_{\text{var}} + \eta_k \|T(g_k, \beta) - T(g^*, \beta)\|_{\text{var}} \\ &\leq (1-\eta_k) \|f_k - f^*\|_{\text{var}} + \eta_k \kappa \|g_k - g^*\|_{\text{var}} \end{aligned}$$

Similarly

$$\|g_{k+1} - f^*\|_{\text{var}} \leq (1-\eta_k) \|g_k - g^*\|_{\text{var}} + \eta_k \kappa \|f_{k+1} - f^*\|_{\text{var}}$$

Therefore

$$\|g_{k+1} - g^*\|_{\text{var}} + \|f_{k+1} - f^*\|_{\text{var}} \leq (1-\eta_k + \kappa^2 \eta_k^2) (\|f_k - f^*\|_{\text{var}} + \|g_k - g^*\|_{\text{var}}),$$

and we still have convergence as long as  $\sum \eta_k = \infty$ . This shows that

$$\frac{f_t + T(g_t, \beta)}{2}, \frac{g_t + T(f_t, \alpha)}{2} \rightarrow f^*, g^*.$$

## 4.2 An adapted mirror descent convergence proof in the non-noisy case

We want to solve

$$\min_{\mu \in \mathcal{M}^+(\mathcal{X}), \nu \in \mathcal{M}^+(\mathcal{X})} F(\mu, \nu) \triangleq \text{KL}(\alpha|\mu) + \text{KL}(\beta|\nu) + \langle \mu \otimes \nu, \exp(-C) \rangle$$

First note that we have

$$D_{F(\mu, \cdot)} = D_{\omega_\alpha(\cdot)} \quad D_{F(\cdot, \nu)} = D_{\omega_\beta(\cdot)},$$

so that at every iteration, we perform a mirror step with a function that is 1-relatively smooth.

Let  $\nu$  be fixed, and let us define  $F_\nu(\cdot) = F(\cdot, \nu)$ . From the relative smoothness of  $F_\nu(\cdot)$  and from its convexity we have, for all  $\mu_x, \mu_y, \mu_z \gg \alpha$ ,

$$\begin{aligned} F(\mu_x, \nu) &\leq F(\mu_y, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_x - \mu_y \rangle + D_{\omega_\alpha}(\mu_x, \mu_y), \\ F(\mu_y, \nu) &\leq F(\mu_z, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_y - \mu_z \rangle \end{aligned}$$

Combining both, we obtain

$$\begin{aligned} F(\mu_x, \nu) &\leq F(\mu_z, \nu) + \langle \nabla_\mu F(\mu_y, \nu), \mu_x - \mu_z \rangle + D_{\omega_\alpha}(\mu_x, \mu_y) \\ \langle \nabla F(\mu_y, \nu), \mu_x - \mu_z \rangle &\geq F(\mu_x, \nu) - F(\mu_z, \nu) - D_{\omega_\alpha}(\mu_x, \mu_y). \end{aligned}$$

We now use the three point property:

$$D_{\omega_\alpha}(\mu_z, \mu_y) - D_{\omega_\alpha}(\mu_z, \mu_x) - D_{\omega_\alpha}(\mu_x, \mu_y) = \langle \nabla \omega_\alpha(\mu_x) - \nabla \omega_\alpha(\mu_y), \mu_z - \mu_x \rangle,$$

replacing  $\mu_y = \mu_k, \mu_x = \mu_{k+1}, \nu = \nu_k$ , we obtain

$$\begin{aligned} D_{\omega_\alpha}(\mu_z, \mu_k) - D_{\omega_\alpha}(\mu_z, \mu_{k+1}) - D_{\omega_\alpha}(\mu_{k+1}, \mu_k) &= \eta_k \langle \nabla_\mu F(\mu_k, \nu_k), \mu_{k+1} - \mu_z \rangle \\ &\geq \eta_k (F(\mu_{k+1}, \nu_k) - F(\mu_z, \nu_k)) - \eta_k D_{\omega_\alpha}(\mu_{k+1}, \mu_k). \end{aligned}$$

Hence, mimicking the derivation for  $\nu$ ,

$$\begin{aligned} \eta_k (F(\mu_{k+1}, \nu_k) - F(\mu_z, \nu_k)) &\leq D_{\omega_\alpha}(\mu_z, \mu_k) - D_{\omega_\alpha}(\mu_z, \mu_{k+1}) - (1 - \eta_k) D_{\omega_\alpha}(\mu_{k+1}, \mu_k) \\ \eta_k (F(\mu_{k+1}, \nu_{k+1}) - F(\mu_{k+1}, \nu_z)) &\leq D_{\omega_\beta}(\nu_z, \nu_k) - D_{\omega_\beta}(\nu_z, \nu_{k+1}) - (1 - \eta_k) D_{\omega_\beta}(\nu_{k+1}, \nu_k) \end{aligned} \quad (2)$$

Setting  $\mu_z = \mu_k, \nu_z = \nu_k$ , we obtain a descent lemma.

$$F(\mu_{k+1}, \nu_{k+1}) \leq F(\mu_{k+1}, \nu_k) \leq F(\mu_k, \nu_k),$$

Summing both equations of (2), we obtain

$$\begin{aligned} \eta_k (F(\mu_{k+1}, \nu_k) + F(\mu_{k+1}, \nu_{k+1}) - (F(\mu_z, \nu_k) + F(\mu_{k+1}, \nu_z))) \\ \leq D_{\omega_\alpha}(\mu_z, \mu_k) + D_{\omega_\beta}(\nu_z, \nu_k) - (D_{\omega_\alpha}(\mu_z, \mu_{k+1}) + D_{\omega_\beta}(\nu_z, \nu_{k+1})) \end{aligned}$$

For  $k \in \mathbb{N}$ , we set  $(\mu_z, \nu_z) = (\mu_k^*, \nu_k^*)$ , such that

$$(\mu_k^*, \nu_k^*) \triangleq \underset{\mu^*, \nu^*}{\operatorname{argmin}} D_{\omega_\beta}(\nu^*, \nu_k) + D_{\omega_\alpha}(\mu^*, \mu_k),$$

and define

$$\begin{aligned} G(\mu_k, \nu_k) &\triangleq \min_{\mu^*, \nu^*} D_{\omega_\beta}(\nu^*, \nu_k) + D_{\omega_\alpha}(\mu^*, \mu_k) \\ &= \text{KL}(\alpha|\mu_k) + \text{KL}(\beta|\nu_k) + 2(\sqrt{\langle \mu^* \otimes \nu, \exp(-C) \rangle \langle \mu \otimes \nu^*, \exp(-C) \rangle} - 1) - F^*. \end{aligned}$$

We obtain

$$\eta_k ((F(\mu_{k+1}, \nu_{k+1}) - F^* \leq G(\mu_k, \nu_k) - G(\mu_{k+1}, \nu_{k+1}) + \eta_k w_k + \eta_k z_k,$$

where

$$\begin{aligned} z_k &= 1 - \langle \mu_{k+1} \otimes \nu_k, \exp(-C) \rangle \\ w_k &= \langle \mu_k^* \otimes \nu_k, \exp(-C) \rangle + \langle \mu_{k+1} \otimes \nu_k^*, \exp(-C) \rangle - 2 \\ &= (\langle \mu_{k+1} \otimes \nu^*, \exp(-C) \rangle \langle \mu^* \otimes \nu_k, \exp(-C) \rangle)^{1/2} \left( \left( \frac{\langle \mu_k \otimes \nu^*, \exp(-C) \rangle}{\langle \mu_{k+1} \otimes \nu^*, \exp(-C) \rangle} \right)^{1/2} + \left( \frac{\langle \mu_{k+1} \otimes \nu^*, \exp(-C) \rangle}{\langle \mu_k \otimes \nu^*, \exp(-C) \rangle} \right)^{1/2} \right) - 2 \end{aligned}$$

The last term is quite ugly, due to the alternated nature of the algorithm.

**Simultaneous updates.** In the non alternated version:

$$\begin{aligned} & \eta_k(F(\mu_{k+1}, \nu_k) + F(\mu_{k+1}, \nu_k) - (F(\mu_z, \nu_k) + F(\mu_k, \nu_z))) \\ & \leq D_{\omega_\alpha}(\mu_z, \mu_k) + D_{\omega_\beta}(\nu_z, \nu_k) - (D_{\omega_\alpha}(\mu_z, \mu_{k+1}) + D_{\omega_\beta}(\nu_z, \nu_{k+1})) \end{aligned}$$