

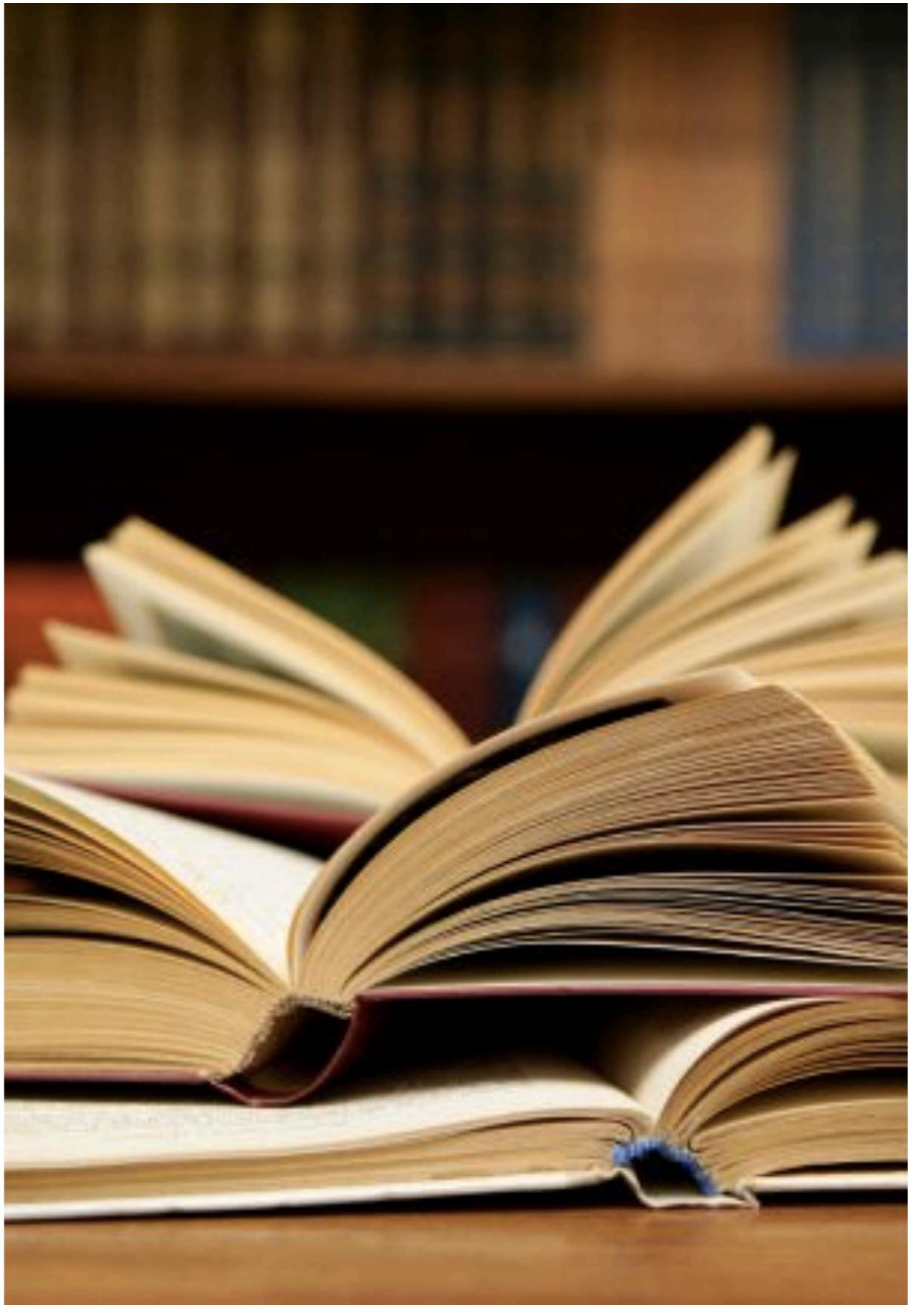
يادگيري برخط

جلسه هجدهم:
پیروی از پیش روی منظم شده
و کاهش آینه‌ای (۱)

فهرست

- بهینه‌سازی بر خط خطی
- الگوریتم‌های جدید بهینه‌سازی بر خط
- بندیت‌سازی
- حل چند مسئله، مثلا بندیت خطی

صورت مسئله



درس یادگیری برخط – ترم پاییز ۱۴۰۰-۱۴۰۱

بهینه‌سازی خطی برخط

$$\mathcal{L} \subset \mathbb{R}^d$$

$$\mathcal{A} \subset \mathbb{R}^d$$

تفاوت با بندیت

بهینه‌سازی برخط غیر

خطی ($l_t(x)$)

۱ - انتخاب عمل

۲ - دریافت

۳ - هزینه: $\langle a_t, y_t \rangle$

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تفاوت با بندیت

بهینه‌سازی برخط غیر

خطی ($l_t(x)$)

۱ - انتخاب عمل

۲ - دریافت

۳ - هزینه: $\langle a_t, y_t \rangle$

$$R_n(a) = \sum_{t=1}^n \langle a_t - a, y_t \rangle$$

$$R_n = \max_{a \in \mathcal{A}} R_n(a)$$

الگوریتم بھینه سازی برخط



کاهش گرادیان برخط

کاهش گرادیان

$$a_{t+1} = a_t - \eta y_t$$

$$a_{t+1} = \Pi(a_t - \eta y_t)$$

کاهش گرادیان تصویر
شده

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$$\arg \min_{a \in A} \|a - (a_t - \eta y_t)\|^2 = \arg \min_{a \in A} (a - (a_t - \eta y_t))^T (a - (a_t - \eta y_t))$$

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کاهش گرادیان تصویر
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$$\begin{aligned}\arg \min_{a \in A} \|a - (a_t - \eta y_t)\|^2 &= \arg \min_{a \in A} (a - (a_t - \eta y_t))^T (a - (a_t - \eta y_t)) \\ &= \arg \min_{a \in A} \|a\|^2 - 2a^T(a_t - \eta y_t) + \|a_t - \eta y_t\|^2\end{aligned}$$

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کاهش گرادیان تصویر شده

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$$a_{t+1} = \arg \min_{a \in A} \eta \langle a, y_t \rangle + D_F(a, a_t)$$

کاہش آپنہ ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

روش‌های پیروی از پیش‌رو

$$a_{t+1} = \arg \min_{a \in A} \sum_s \langle a, y_s \rangle$$

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تابع لژاندر

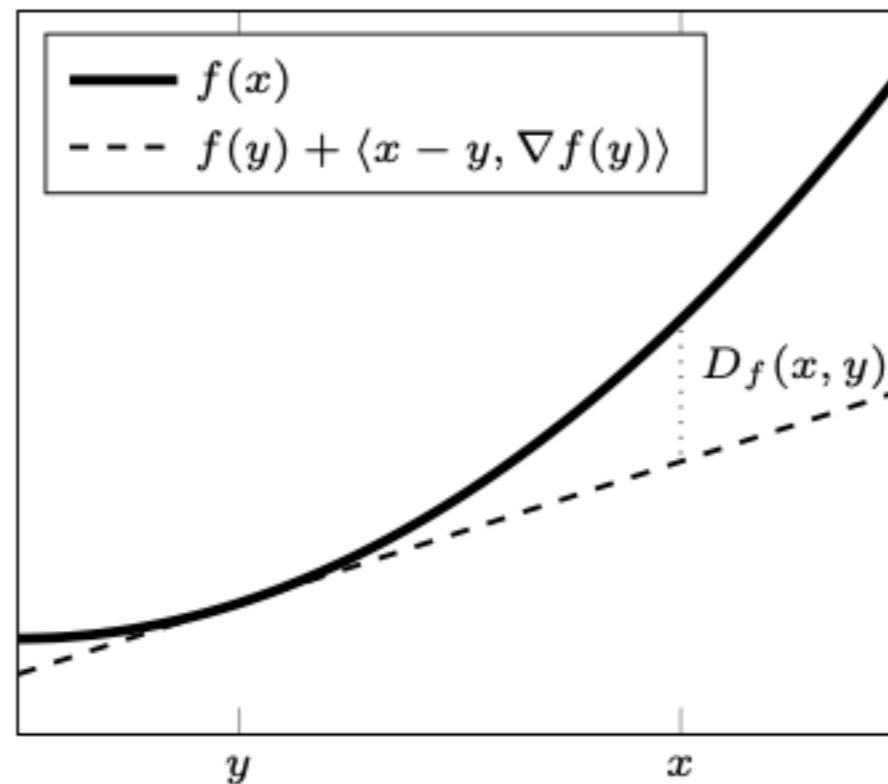
لژاندر: مانند بشقاب ته گود

- (a) C is non-empty;
- (b) f is differentiable and strictly convex on C ; and
- (c) $\lim_{n \rightarrow \infty} \|\nabla f(x_n)\|_2 = \infty$ for any sequence $(x_n)_n$ with $x_n \in C$ for all n and $\lim_{n \rightarrow \infty} x_n = x$ and some $x \in \partial C$.

فاصله D_f

لړاندې: مانند بشقاب ته ګود

$$D_f(x, y) = f(x) - f(y) - \langle x - y, \nabla f(y) \rangle$$



چند تابع خوب لرماندر

$$f(x) = \frac{1}{2} \|x\|_2^2 \quad \text{dom}(f) = \mathbb{R}^d$$

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$$f(x) = \sum_i x_i \log(x_i) - x_i \quad \text{dom}(f) = [0, \infty)^d$$

برای توزیع x و y

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برای توزیع x و y

$$D_f(x, y) = f(x) - f(y) - \langle x - y, \nabla f(y) \rangle$$

$$= \langle x, \log(x) \rangle - \langle x, 1 \rangle - \langle y, \log(y) \rangle + \langle y, 1 \rangle - \langle x, \log(y) \rangle + \langle y, \log(y) \rangle$$

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$$= \sum x_i \log(x_i / y_i)$$

دو الگوریتم

کاهش آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

پیروی از پیش‌روی
منظلم شده

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مثال: کاہش آینه‌ای

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آینه‌ای

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$$\nabla(\dots) = \eta y_t + a - a_t = 0$$

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کاهش گرادیان برخط

مثال: کاهش آینه‌ای

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کاهش گرادیان برخط

مثال: کاہش آینه‌ای

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$$a_{t+1} = \Pi(a_t - \eta y_t)$$

کاهش گرادیان برخط
تصویر شده

مثال: کاهش آینه‌ای

$$A = \mathbb{R}^d \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \nabla F(a) = a \quad D(a, a_t) = \frac{1}{2}\|a - a_t\|_2^2$$

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$$\nabla(\dots) = \eta y_t + a - a_t = 0$$

$$a_{t+1} = a_t - \eta y_t$$

کاهش گرادیان برخط

محدب =

کاهش گرادیان برخط تصویرشده

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \eta \langle a, y_t \rangle + \frac{1}{2}\|a - a_t\|_2^2 = \Pi(a_t - \eta y_t)$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

$$A = \mathbb{R}^d \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \nabla F(a) = a \quad D(a, a_t) = \frac{1}{2} \|a - a_t\|_2^2$$

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$$a_{t+1} = \arg \min_a \eta \sum_s \langle a, y_s \rangle + \frac{1}{2} \|a\|^2$$

$$\nabla(\dots) = \eta \sum y_s + a = 0$$

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منظم شده

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$$A = \mathbb{R}^d \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \nabla F(a) = a \quad D(a, a_t) = \frac{1}{2} \|a - a_t\|_2^2$$

$$a_{t+1} = \arg \min_a \eta \sum_s \langle a, y_s \rangle + \frac{1}{2} \|a\|^2$$

$$\nabla(\dots) = \eta \sum_s y_s + a = 0$$

$$a_{t+1} = -\eta \sum_{s=1}^t y_s$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم شده

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$$a_{t+1} = -\eta \sum_{s=1}^t y_s = a_t - \eta y_t$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم شده

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$$\nabla(\dots) = \eta \sum_s y_s + a = 0$$

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کاهش گرادیان برخط

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم
شده

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$$A = \mathcal{P}_{d-1} \quad f(x) = \sum_i x_i \log(x_i) - x_i \quad \text{dom}(f) = [0, \infty)^d$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم
شده

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قضیه: در شرایط خوب با تابع لژاندر f

$$y = \operatorname{argmin}_{z \in A} f(z)$$

$$\tilde{y} = \operatorname{argmin}_{z \in \mathbb{R}^d} f(z)$$

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$$D_f(x, y) = f(x) - f(y) - \langle x - y, \nabla f(y) \rangle$$

$$= \langle x, \log(x) \rangle - \langle x, 1 \rangle - \langle y, \log(y) \rangle + \langle y, 1 \rangle - \langle x, \log(y) \rangle + \langle y, \log(y) \rangle$$

$$= \sum x_i \log(x_i/y_i) + C$$

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$$\nabla \sum_i x_i \log(x_i/\tilde{y}_i) = \alpha 1$$

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$$x_i \propto \tilde{y}_i$$

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$$a_{t+1} = \arg \min_{a \in \mathcal{P}_{d-1}} \sum_i x_i \log(x_i/\tilde{y}_i)$$

$$\nabla \sum_i x_i \log(x_i/\tilde{y}_i) = \alpha 1 \quad \log(x_i) + 1 - \log(\tilde{y}_i) = \alpha$$

$$\log(x_i/\tilde{y}_i) = \alpha - 1$$

$$x_i \propto \tilde{y}_i$$

$$x_i = \tilde{y}_i / \|\tilde{y}\|$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم
شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

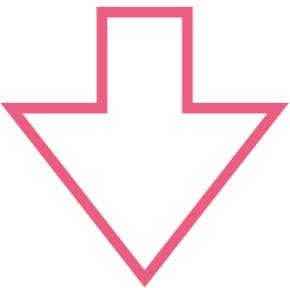
$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

$$A = \mathcal{P}_{d-1} \quad f(x) = \sum_i x_i \log(x_i) - x_i \quad \text{dom}(f) = [0, \infty)^d$$

$$a_{t+1,i} = \frac{\exp\left(-\eta \sum_{s=1}^t y_{si}\right)}{\sum_{j=1}^d \exp\left(-\eta \sum_{s=1}^t y_{sj}\right)}.$$

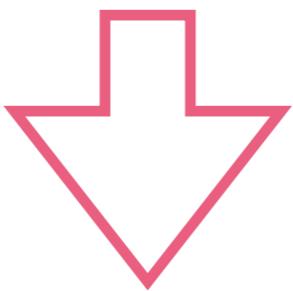
عملیات محاسبه کاٹپنے آئندہ ایجاد

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

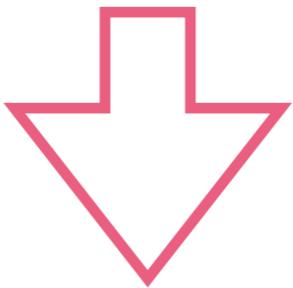


عملیات محاسبه کا مقتضی اینہے ایسا

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$



$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$



$$\tilde{a}_{t+1} = \operatorname{argmin}_{a \in \mathcal{D}} \eta \langle a, y_t \rangle + D_F(a, a_t)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} D_F(a, \tilde{a}_{t+1}).$$

يادگيري برخط

جلسه نوزدهم:
پیروی از پیش روی منظم شده
و کاهش آینه‌ای (۲)

مرور

- بهینه‌سازی بر خط خطی
- الگوریتم‌های جدید بهینه‌سازی بر خط
- بندیت‌سازی
- حل چند مسئله، مثلا بندیت خطی

تابع لژاندر

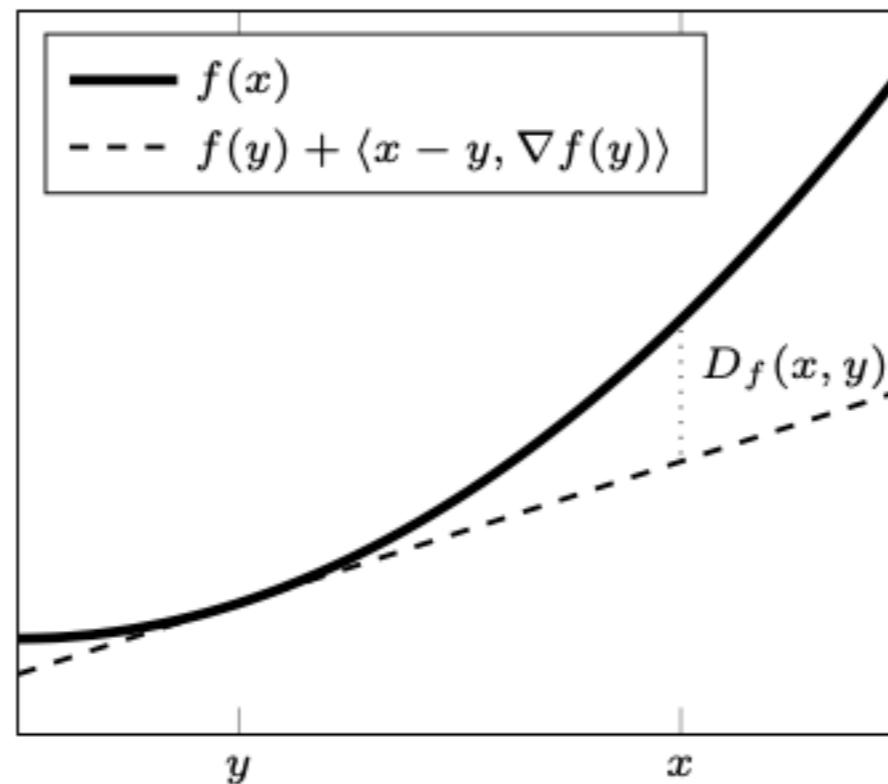
لژاندر: مانند بشقاب ته گود

- (a) C is non-empty;
- (b) f is differentiable and strictly convex on C ; and
- (c) $\lim_{n \rightarrow \infty} \|\nabla f(x_n)\|_2 = \infty$ for any sequence $(x_n)_n$ with $x_n \in C$ for all n and $\lim_{n \rightarrow \infty} x_n = x$ and some $x \in \partial C$.

فاصله D_f

لړاندې: مانند بشقاب ته ګود

$$D_f(x, y) = f(x) - f(y) - \langle x - y, \nabla f(y) \rangle$$



دو الگوریتم

کاهش آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

پیروی از پیش‌روی
منظلم شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

مثال: کاهش آینه‌ای

$$A = \mathbb{R}^d \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \nabla F(a) = a \quad D(a, a_t) = \frac{1}{2}\|a - a_t\|_2^2$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathbb{R}^d} \eta \langle a, y_t \rangle + \frac{1}{2}\|a - a_t\|_2^2$$

$$\nabla(\dots) = \eta y_t + a - a_t = 0$$

$$a_{t+1} = a_t - \eta y_t$$

کاهش گرادیان برخط

محدب =

کاهش گرادیان برخط تصویرشده

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \eta \langle a, y_t \rangle + \frac{1}{2}\|a - a_t\|_2^2 = \Pi(a_t - \eta y_t)$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم
شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

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$$a_{t+1,i} = \frac{\exp\left(-\eta \sum_{s=1}^t y_{si}\right)}{\sum_{j=1}^d \exp\left(-\eta \sum_{s=1}^t y_{sj}\right)}.$$

قضیه: در شرایط خوب با تابع لژاندر f

$$y = \operatorname{argmin}_{z \in A} f(z)$$

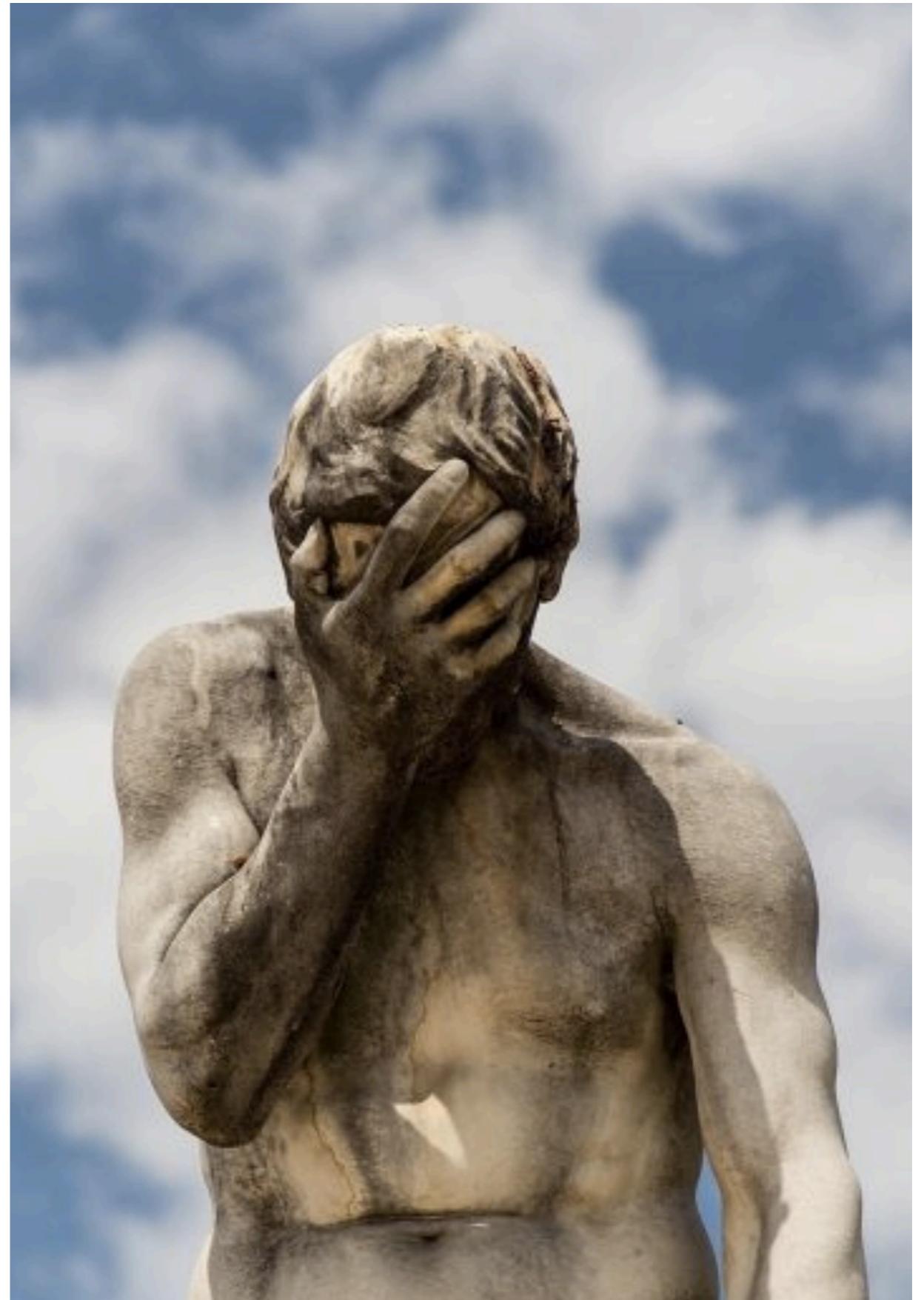
$$\tilde{y} = \operatorname{argmin}_{z \in \mathbb{R}^d} f(z)$$

$$y = \operatorname{argmin}_{z \in A} D_f(z, \tilde{y}).$$

تحلیل پژیمانی

کاهش آینه‌ای

پیروی از پیش‌روی منظم شده



تحلیل پیشمانی کاہش آینه‌ای

کاہش
آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

قضیه:

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$\tilde{a}_{t+1} = \operatorname{argmin}_{a \in \mathcal{D}} \eta \langle a, y_t \rangle + D_F(a, a_t)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} D_F(a, \tilde{a}_{t+1}).$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$(\mu-\mu_0)^2\leq \frac{1}{2}\left(\frac{\partial^2 F}{\partial x^2}(x_0)+\frac{\partial^2 F}{\partial x^2}(x_1)\right)$$

$$\langle a-a_{t+1}, \eta y_t+\nabla F(a_{t+1})-\nabla F(a_t) \rangle \geq 0$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a-a_{t+1},\eta y_t+\nabla F(a_{t+1})-\nabla F(a_t)\rangle\geq 0$$

$$a_{t+1}=\operatorname{argmin}_{a\in\mathcal{A}}\left(\eta\langle a,y_t\rangle+D_F(a,a_t)\right)$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

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$$a_{t+1}=\operatorname{argmin}_{a\in\mathcal{A}}\left(\eta\langle a,y_t\rangle+D_F(a,a_t)\right)$$

$$\langle a_{t+1}-a,y_t\rangle\leq\frac{1}{\eta}\langle a-a_{t+1},\nabla F(a_{t+1})-\nabla F(a_t)\rangle$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a-a_{t+1},\eta y_t+\nabla F(a_{t+1})-\nabla F(a_t)\rangle\geq 0$$

$$a_{t+1}=\operatorname{argmin}_{a\in\mathcal{A}}\left(\eta\langle a,y_t\rangle+D_F(a,a_t)\right)$$

$$\begin{aligned}\langle a_{t+1}-a, y_t\rangle &\leq \frac{1}{\eta}\langle a-a_{t+1}, \nabla F(a_{t+1})-\nabla F(a_t)\rangle \\&= \frac{1}{\eta}\left(D(a,a_t)-D(a,a_{t+1})-D(a_{t+1},a_t)\right)\\F(a)&&-F(a)&&-F(a_{t+1})\\-F(a_t)&&+F(a_{t+1})&&+F(a_t)\\-\langle a-a_t,\nabla F(a_t)\rangle&&+\langle a-a_{t+1},\nabla F(a_{t+1})\rangle&&+\langle a_{t+1}-a,\nabla F(a_t)\rangle\end{aligned}$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a - a_{t+1}, \eta y_t + \nabla F(a_{t+1}) - \nabla F(a_t) \rangle \geq 0$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \langle a, y_t \rangle + D_F(a, a_t) \right)$$

$$\langle a_{t+1} - a, y_t \rangle \leq \frac{1}{\eta} \langle a - a_{t+1}, \nabla F(a_{t+1}) - \nabla F(a_t) \rangle$$

$$= \frac{1}{\eta} \left(D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t) \right)$$

$$\cancel{F(a)} \hspace{10em} \cancel{-F(a)} \hspace{10em} \cancel{-F(a_{t+1})}$$

$$-F(a_t) \hspace{10em} +F(a_{t+1}) \hspace{10em} +F(a_t)$$

$$-\langle a - a_t, \nabla F(a_t) \rangle \quad +\langle a - a_{t+1}, \nabla F(a_{t+1}) \rangle \quad +\langle a_{t+1} - a_t, \nabla F(a_t) \rangle$$

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a - a_{t+1}, \eta y_t + \nabla F(a_{t+1}) - \nabla F(a_t) \rangle \geq 0$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

$$\langle a_{t+1} - a, y_t \rangle \leq \frac{1}{\eta} \langle a - a_{t+1}, \nabla F(a_{t+1}) - \nabla F(a_t) \rangle$$

$$= \frac{1}{\eta} (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t))$$

$F(\cancel{a})$	$-\cancel{F}(a)$	$-\cancel{F}(a_{t+1})$
-----------------	------------------	------------------------

$-F(a_t)$	$+F(\cancel{a}_{t+1})$	$+F(a_t)$
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$-\langle a - a_t, \nabla F(a_t) \rangle$	$+\langle a - a_{t+1}, \nabla F(a_{t+1}) \rangle$	$+\langle a_{t+1} - a_t, \nabla F(a_t) \rangle$
---	---	---

$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

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$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

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$$= \frac{1}{\eta} (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t))$$

$F(\cancel{a})$	$-\cancel{F}(a)$	$-\cancel{F}(a_{t+1})$
-----------------	------------------	------------------------

$-\cancel{F}(\cancel{a}_t)$	$+\cancel{F}(a_{t+1})$	$+F(\cancel{a}_t)$
-----------------------------	------------------------	--------------------

$-\langle a - a_t, \nabla F(a_t) \rangle$	$+\langle a - a_{t+1}, \nabla F(a_{t+1}) \rangle$	$+\langle a_{t+1} - a_t, \nabla F(a_t) \rangle$
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$$\langle a_t - a, y_t \rangle = \langle a_t - a_{t+1}, y_t \rangle + \langle a_{t+1} - a, y_t \rangle$$

$$\langle a - a_{t+1}, \eta y_t + \nabla F(a_{t+1}) - \nabla F(a_t) \rangle \geq 0$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

$$\langle a_{t+1} - a, y_t \rangle \leq \frac{1}{\eta} \langle a - a_{t+1}, \nabla F(a_{t+1}) - \nabla F(a_t) \rangle$$

$$= \frac{1}{\eta} (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t))$$

$F(\cancel{a})$	$-F(\cancel{a})$	$-F(\cancel{a}_{t+1})$
-----------------	------------------	------------------------

$-F(\cancel{a}_t)$	$+F(\cancel{a}_{t+1})$	$+F(\cancel{a}_t)$
--------------------	------------------------	--------------------

$-\langle a - \cancel{a}_t, \nabla F(a_t) \rangle$	$+\langle a - a_{t+1}, \nabla F(a_{t+1}) \rangle$	$+\langle a_{t+1} - \cancel{a}_t, \nabla F(a_t) \rangle$
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$$\begin{aligned}
R_n &= \sum_{t=1}^n \langle a_t - a, y_t \rangle \\
&\leq \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{1}{\eta} \sum_{t=1}^n (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t))
\end{aligned}$$

$$\begin{aligned}
R_n &= \sum_{t=1}^n \langle a_t - a, y_t \rangle \\
&\leq \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{1}{\eta} \sum_{t=1}^n (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t)) \\
&= \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{1}{\eta} \left(D(a, a_1) - D(a, a_{n+1}) - \sum_{t=1}^n D(a_{t+1}, a_t) \right)
\end{aligned}$$

$$\begin{aligned}
R_n &= \sum_{t=1}^n \langle a_t - a, y_t \rangle \\
&\leq \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{1}{\eta} \sum_{t=1}^n (D(a, a_t) - D(a, a_{t+1}) - D(a_{t+1}, a_t)) \\
&= \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{1}{\eta} \left(D(a, a_1) - D(a, a_{n+1}) - \sum_{t=1}^n D(a_{t+1}, a_t) \right) \\
&\leq \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle + \frac{F(a) - F(a_1)}{\eta} - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t)
\end{aligned}$$

$$\eta y_t = \nabla F(a_t) - \nabla F(a_{t+1})$$

$$\langle a_t - a_{t+1}, y_t \rangle = \frac{1}{\eta} \langle a_t - a_{t+1}, \nabla F(a_t) - \nabla F(a_{t+1}) \rangle$$

$$\eta y_t = \nabla F(a_t) - \nabla F(a_{t+1})$$

$$\begin{aligned}\langle a_t - a_{t+1}, y_t \rangle &= \frac{1}{\eta} \langle a_t - a_{t+1}, \nabla F(a_t) - \nabla F(a_{t+1}) \rangle \\ &= \frac{1}{\eta} (D(a_{t+1}, a_t) + D(a_t, \tilde{a}_{t+1}) - D(a_{t+1}, \tilde{a}_{t+1}))\end{aligned}$$

$$\eta y_t = \nabla F(a_t) - \nabla F(a_{t+1})$$

$$\begin{aligned}\langle a_t - a_{t+1}, y_t \rangle &= \frac{1}{\eta} \langle a_t - a_{t+1}, \nabla F(a_t) - \nabla F(a_{t+1}) \rangle \\ &= \frac{1}{\eta} (D(a_{t+1}, a_t) + D(a_t, \tilde{a}_{t+1}) - D(a_{t+1}, \tilde{a}_{t+1})) \\ &\leq \frac{1}{\eta} (D(a_{t+1}, a_t) + D(a_t, \tilde{a}_{t+1}))\end{aligned}$$

تحلیل پیشمانی پیروی منظم شده

پیشمانی
منظم
شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

قضیه:

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

کاهش آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

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کاهش گرادیان تصویر شده

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

قضیه:

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$\tilde{a}_{t+1} = \operatorname{argmin}_{a \in \mathcal{D}} \eta \langle a, y_t \rangle + D_F(a, a_t)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} D_F(a, \tilde{a}_{t+1}).$$

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

مثال (گوی واحد)

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$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2$$

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

مثال (گوی واحد)

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$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2$$

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2 \leq \frac{1}{2\eta} + \frac{\eta n}{2}$$

مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2 \leq \frac{1}{2\eta} + \frac{\eta n}{2} = \sqrt{n}.$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

$$R_n \leq O(\sqrt{nD}G)$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$R_n \leq O(\sqrt{nD}G)$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2$$

$$R_n \leq O(\sqrt{nD}G)$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

$$R_n \leq O(\sqrt{nD}G)$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2}\|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2}\|y_t\|_2^2$$

$$R_n \leq O(\sqrt{nDG})$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2}\|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2}\|y_t\|_2^2$$

$$R_n \leq O(\sqrt{nDG})$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2 \leq \frac{1}{2\eta} + \frac{\eta n}{2}$$

مثال (گوی واحد) ==> حالت کلی

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \cdot \text{آنگاه}$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$D(a_t, \tilde{a}_{t+1}) = \frac{1}{2} \|\tilde{a}_{t+1} - a_t\|_2^2 = \frac{\eta^2}{2} \|y_t\|_2^2$$

$$R_n \leq O(\sqrt{nD}G)$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_2^2 \leq \frac{1}{2\eta} + \frac{\eta n}{2} = \sqrt{n}.$$

مثال: کاهش آینه‌ای روی سادک

کاهش
آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

$$A = \mathcal{P}_{d-1} \quad f(x) = \sum_i x_i \log(x_i) - x_i \quad \text{dom}(f) = [0, \infty)^d$$

$$y_t \in \mathcal{L} = [0, 1]^d$$

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مثال: کاهش آینه‌ای روی سادک

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مثال: کاهش آینه‌ای روی سادک

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آینه‌ای

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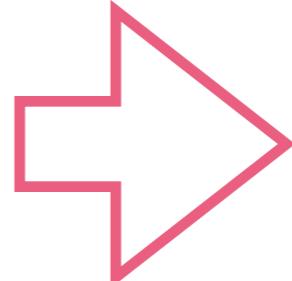
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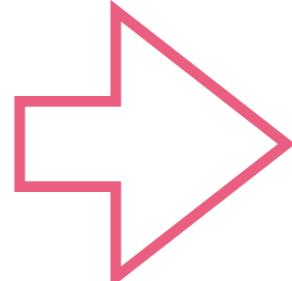
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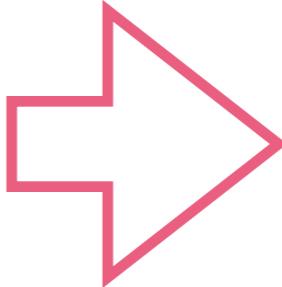
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$$a_{t+1} = \frac{a_t \exp(-\eta y_t)}{\|a_t \exp(-\eta y_t)\|_1}$$

مثال کاہش آپنه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

مثال کاہش آپنه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$\leq \boxed{\frac{\log(d)}{\eta}}$$

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$$\leq \boxed{\frac{\log(d)}{\eta}} + \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty}$$

مثال کاهش آینه‌ای روی سادک

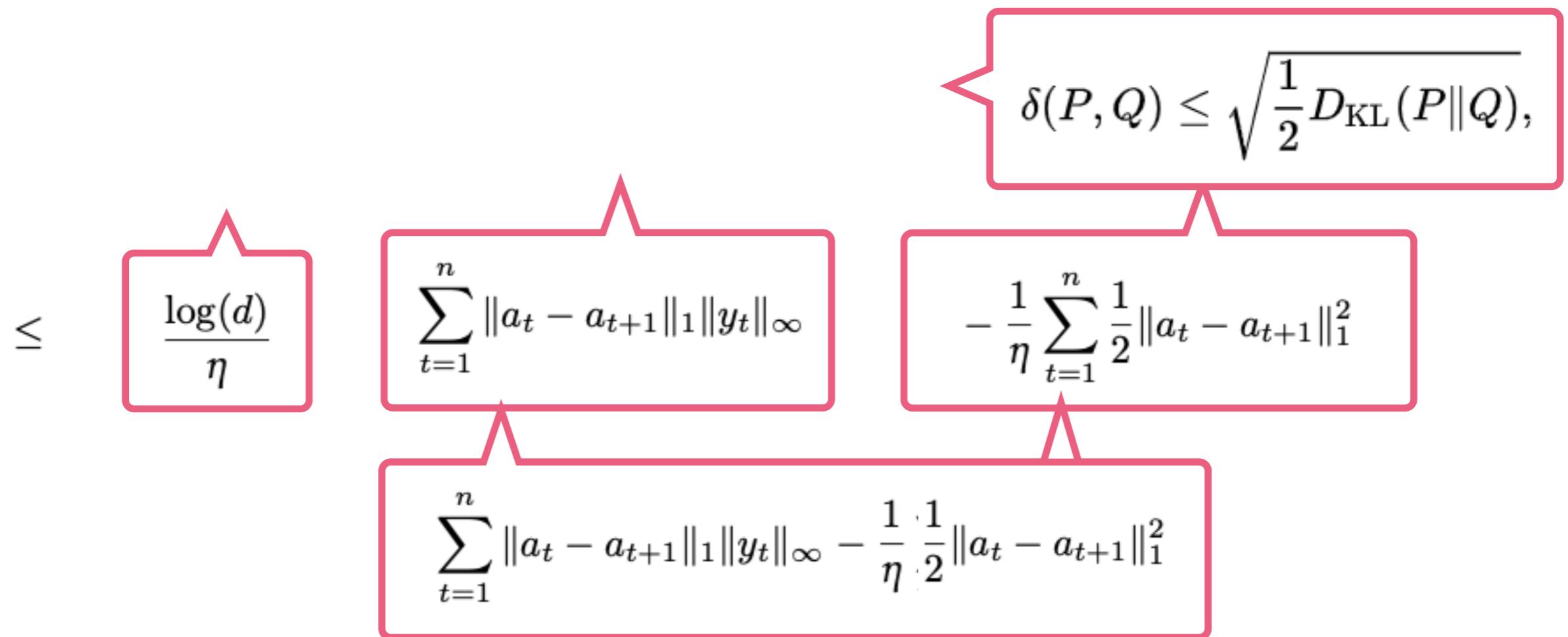
$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$\leq \boxed{\frac{\log(d)}{\eta}} + \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty} - \boxed{-\frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2}$$

$\delta(P, Q) \leq \sqrt{\frac{1}{2} D_{\text{KL}}(P\|Q)},$

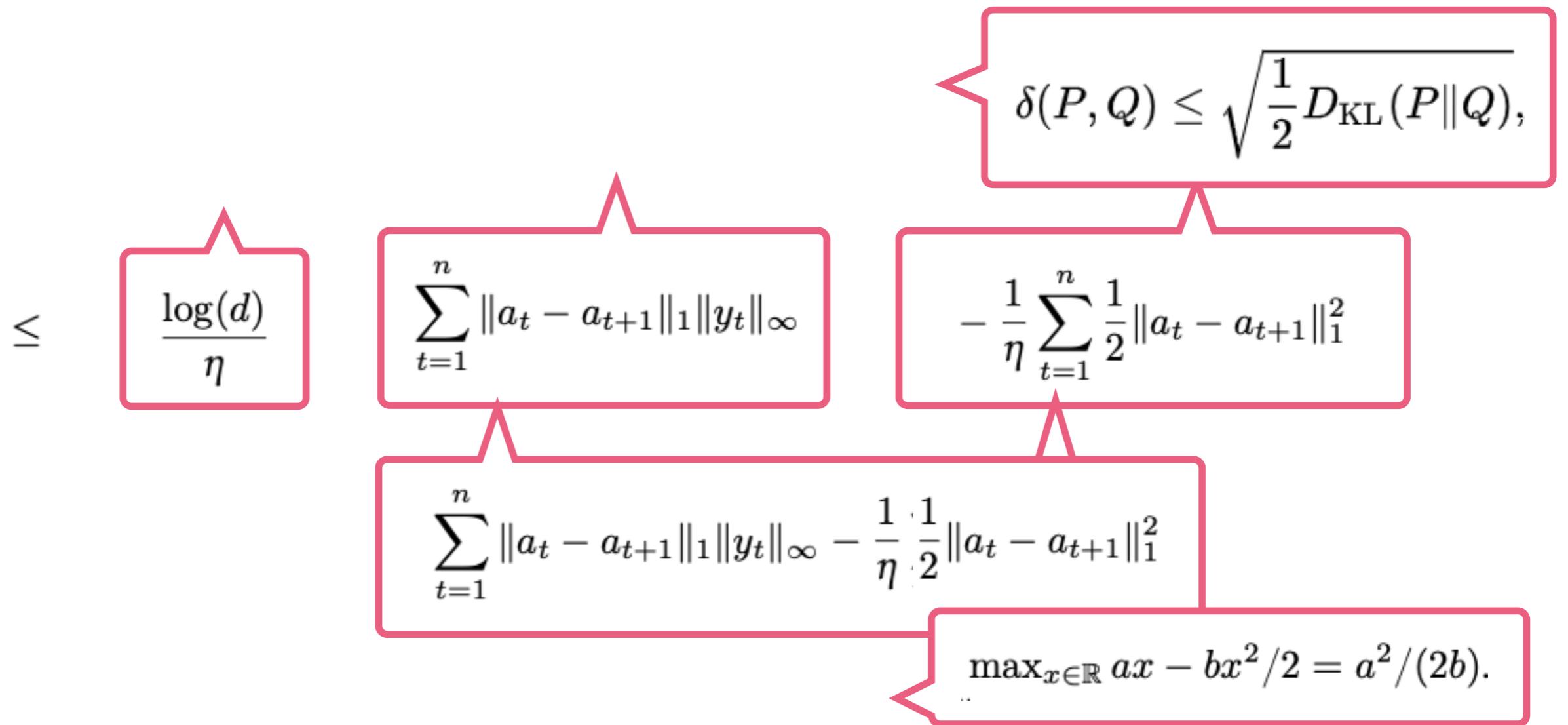
مثال کاهش آینه‌ای روی سادک

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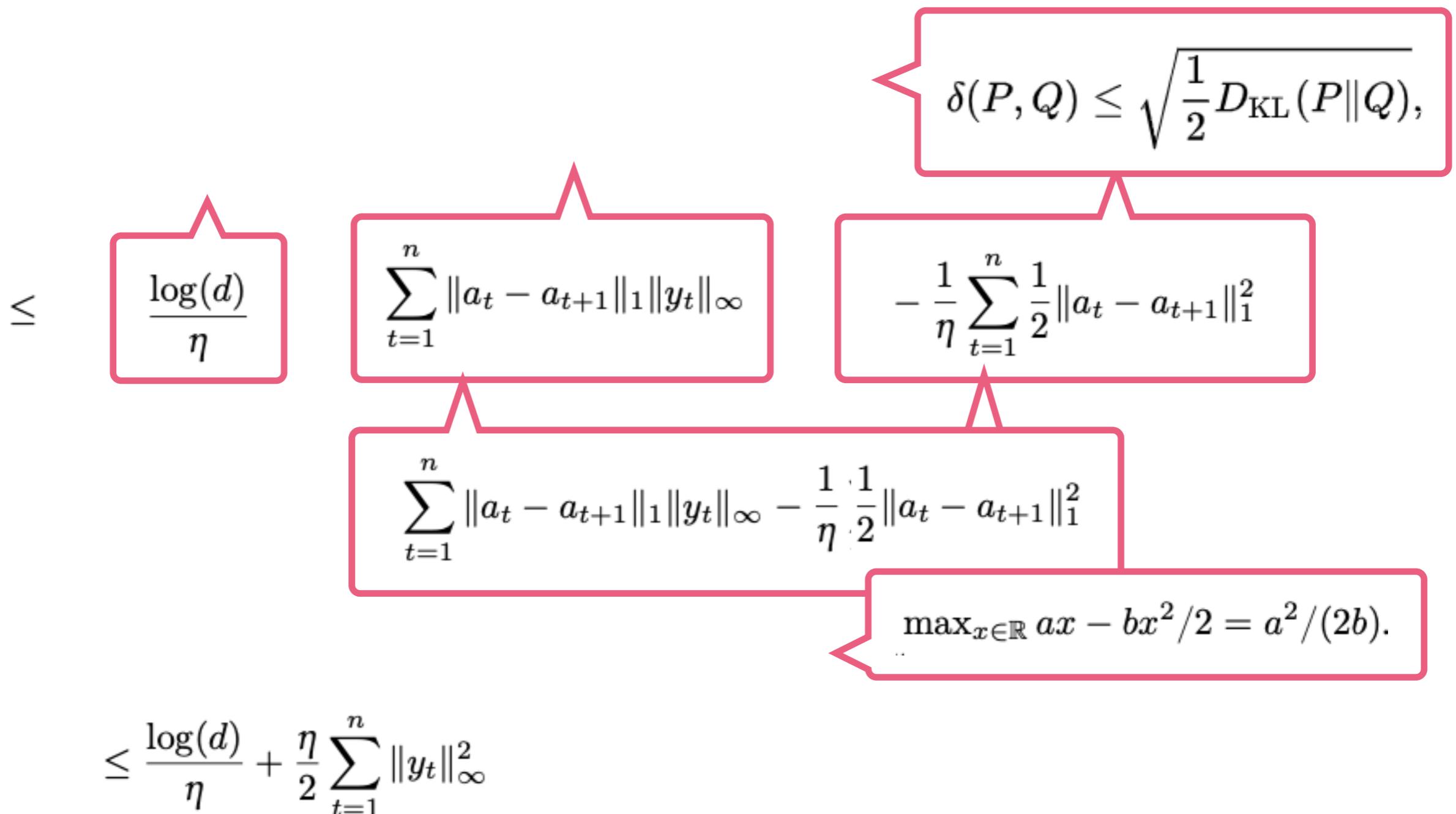
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$$\leq \frac{\log(d)}{\eta} + \sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2$$

$$\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \cdot \frac{1}{2} \|a_t - a_{t+1}\|_1^2$$

$$\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).$$

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 & \quad \leftarrow \boxed{\frac{\log(d)}{\eta}} \quad \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty} \quad \boxed{- \frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2} \\
 & \quad \leftarrow \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \cdot \frac{1}{2} \|a_t - a_{t+1}\|_1^2} \\
 & \quad \leftarrow \boxed{\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).}
 \end{aligned}$$

$$\leq \frac{\log(d)}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_\infty^2 \leq \frac{\log(d)}{\eta} + \frac{\eta n}{2} = \sqrt{2n \log(d)}$$

مثال: کاهش آینه‌ای روی سادک

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مثال: کاهش آینه‌ای روی سادک

کاهش
آینه‌ای

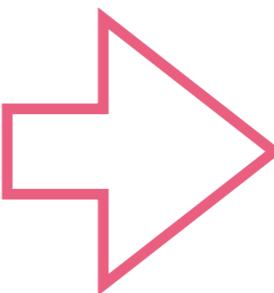
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مثال: کاهش آینه‌ای روی سادک

کاهش
آینه‌ای

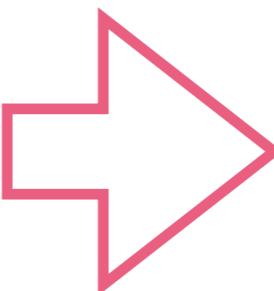
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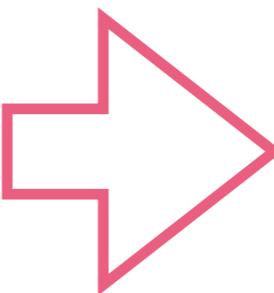
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$$a_{t+1} = \Pi(\tilde{a}_{t+1})$$

مثال کاهش آینه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

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 & \leq \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\| \|y_t\|} \\
 & \quad \boxed{- \frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2} \\
 & \leq \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \frac{1}{2} \|a_t - a_{t+1}\|_1^2} \\
 & \quad \boxed{\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).}
 \end{aligned}$$

$$\leq \frac{\log(d)}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_\infty^2 \leq \frac{\log(d)}{\eta} + \frac{\eta n}{2} = \sqrt{2n \log(d)}$$

مثال کاهش آینه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$\delta(P, Q) \leq \sqrt{\frac{1}{2} D_{\text{KL}}(P \| Q)},$$

$$\leq \boxed{\frac{1}{\eta}}$$

$$\sum_{t=1}^n \|a_t - a_{t+1}\| \boxed{\|y_t\|}$$

$$-\frac{1}{\eta} \sum_{t=1}^n \boxed{\frac{1}{2} \|a_t - a_{t+1}\|_1^2}$$

$$\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \cdot \frac{1}{2} \|a_t - a_{t+1}\|_1^2$$

$$\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).$$

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 \end{aligned}$$

$$\leq \boxed{\frac{1}{\eta}} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|^2 \leq \boxed{\frac{1}{\eta}} + \frac{\eta n}{2} d = \sqrt{2n \log(d)}$$

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 & \quad \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\| \|y_t\| - \frac{1}{\eta} \frac{1}{2} \|a_t - a_{t+1}\|^2} \\
 & \quad \boxed{\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).} \\
 & \leq \boxed{\frac{1}{\eta}} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|^2 \leq \boxed{\frac{1}{\eta}} + \frac{\eta n}{2} d = \boxed{O(\sqrt{nd})} \\
 & \quad \boxed{d = \sqrt{2n \log(d)}}
 \end{aligned}$$

يادگيري برخط

جلسه بیستم:

پیروی از پیش روی منظم شده
و کاهش آینه‌ای (۳)

مرور

- بهینه‌سازی بر خط خطی
- الگوریتم‌های جدید بهینه‌سازی بر خط
- بندیت‌سازی
- حل چند مسئله، مثلا بندیت خطی

تابع لژاندر

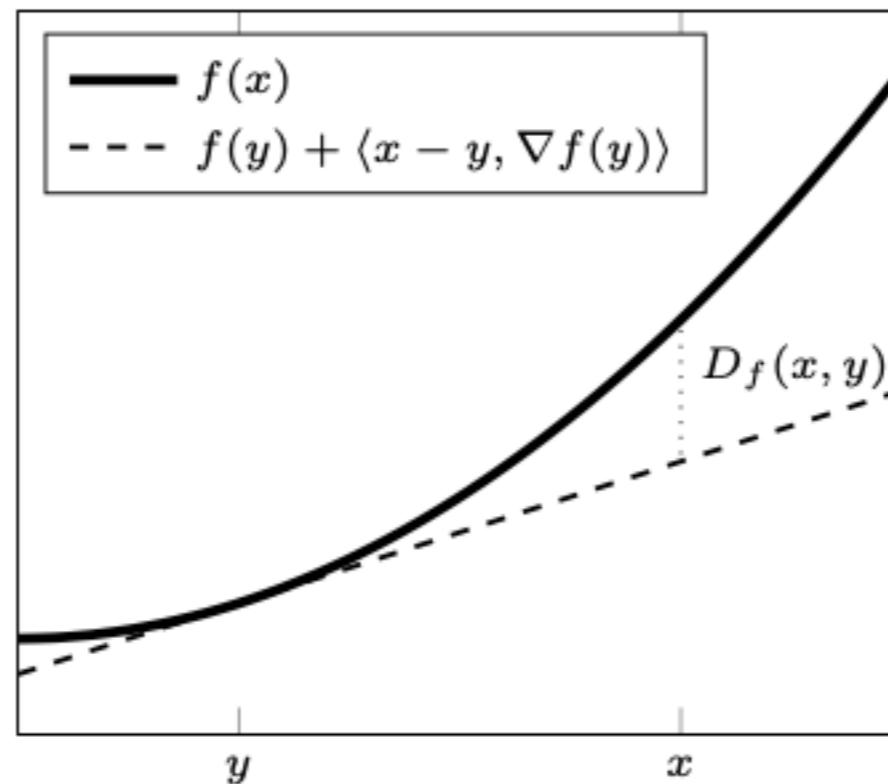
لژاندر: مانند بشقاب ته گود

- (a) C is non-empty;
- (b) f is differentiable and strictly convex on C ; and
- (c) $\lim_{n \rightarrow \infty} \|\nabla f(x_n)\|_2 = \infty$ for any sequence $(x_n)_n$ with $x_n \in C$ for all n and $\lim_{n \rightarrow \infty} x_n = x$ and some $x \in \partial C$.

فاصله D_f

لړاندې: مانند بشقاب ته ګود

$$D_f(x, y) = f(x) - f(y) - \langle x - y, \nabla f(y) \rangle$$



دو الگوریتم

کاهش آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

پیروی از پیش‌روی
منظلم شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

مثال: کاهش آینه‌ای

$$A = \mathbb{R}^d \quad F(a) = \frac{1}{2}\|a\|_2^2 \quad \nabla F(a) = a \quad D(a, a_t) = \frac{1}{2}\|a - a_t\|_2^2$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathbb{R}^d} \eta \langle a, y_t \rangle + \frac{1}{2}\|a - a_t\|_2^2$$

$$\nabla(\dots) = \eta y_t + a - a_t = 0$$

$$a_{t+1} = a_t - \eta y_t$$

کاهش گرادیان برخط

محدب =

کاهش گرادیان برخط تصویرشده

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \eta \langle a, y_t \rangle + \frac{1}{2}\|a - a_t\|_2^2 = \Pi(a_t - \eta y_t)$$

مثال: پیروی از پیش روی منظم شده

پیش روی
منظم
شده

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

$$A = \mathcal{P}_{d-1} \quad f(x) = \sum_i x_i \log(x_i) - x_i \quad \operatorname{dom}(f) = [0, \infty)^d$$

$$a_{t+1,i} = \frac{\exp\left(-\eta \sum_{s=1}^t y_{si}\right)}{\sum_{j=1}^d \exp\left(-\eta \sum_{s=1}^t y_{sj}\right)}.$$

قضیه: در شرایط خوب با تابع لژاندر f

$$y = \operatorname{argmin}_{z \in A} f(z)$$

$$\tilde{y} = \operatorname{argmin}_{z \in \mathbb{R}^d} f(z)$$

$$y = \operatorname{argmin}_{z \in A} D_f(z, \tilde{y}).$$

تحلیل پیشمانی کاہش آینه‌ای

کاہش
آینه‌ای

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

قضیه:

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

$$R_n(a) \leq \frac{1}{\eta} \left(F(a) - F(a_1) + \sum_{t=1}^n D(a_t, \tilde{a}_{t+1}) \right)$$

$$\tilde{a}_{t+1} = \operatorname{argmin}_{a \in \mathcal{D}} \eta \langle a, y_t \rangle + D_F(a, a_t)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} D_F(a, \tilde{a}_{t+1}).$$

تحلیل پیشمانی پیروی منظم شده

پیشمانی
منظم
شده

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مثال (گوی واحد)

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\eta = \sqrt{1/n} \quad F(a) = \frac{1}{2} \|a\|_2^2 \quad \text{الگوریتم کاهش آینه‌ای}$$

$$R_n \leq \sqrt{n} \quad \text{آنگاه}$$

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مثال کاہش آپنه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

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$$\leq \boxed{\frac{\log(d)}{\eta}}$$

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$$\leq \boxed{\frac{\log(d)}{\eta}} + \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty}$$

مثال کاهش آینه‌ای روی سادک

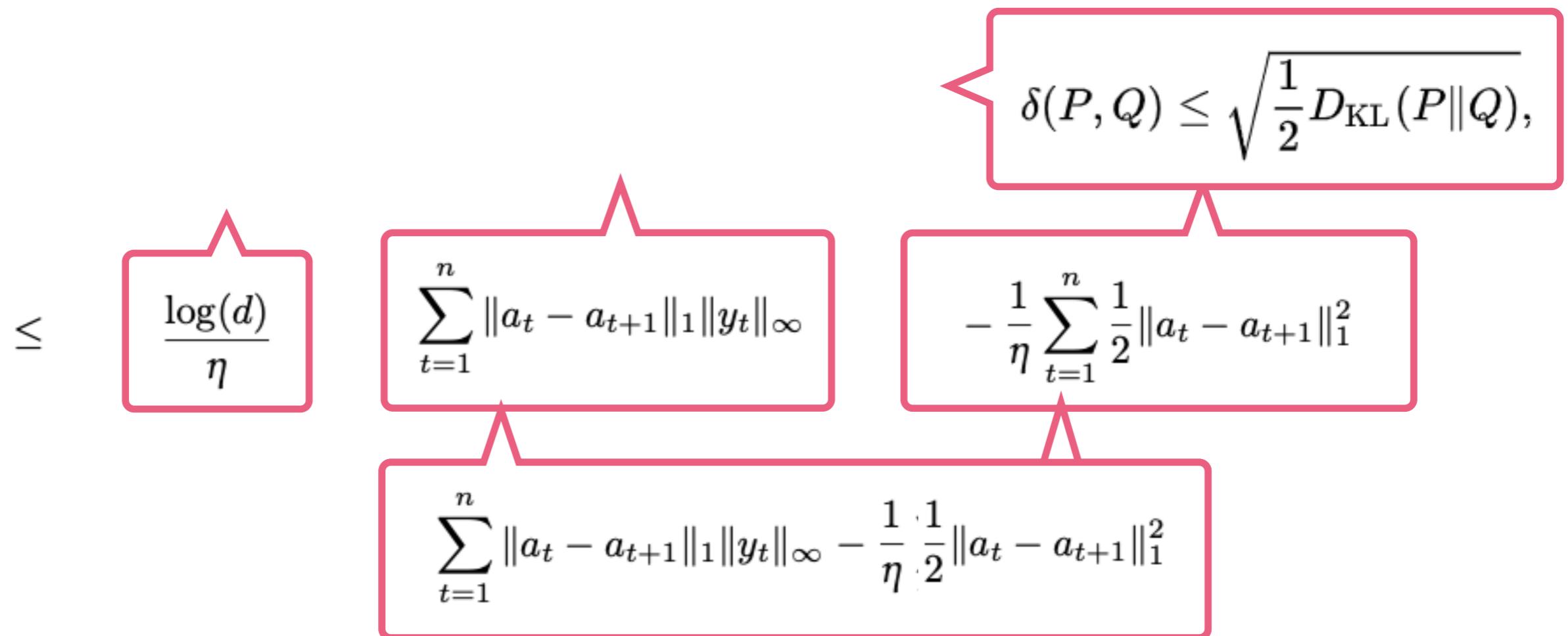
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$$\leq \boxed{\frac{\log(d)}{\eta}} + \boxed{\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty} - \boxed{-\frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2}$$

$\delta(P, Q) \leq \sqrt{\frac{1}{2} D_{\text{KL}}(P\|Q)},$

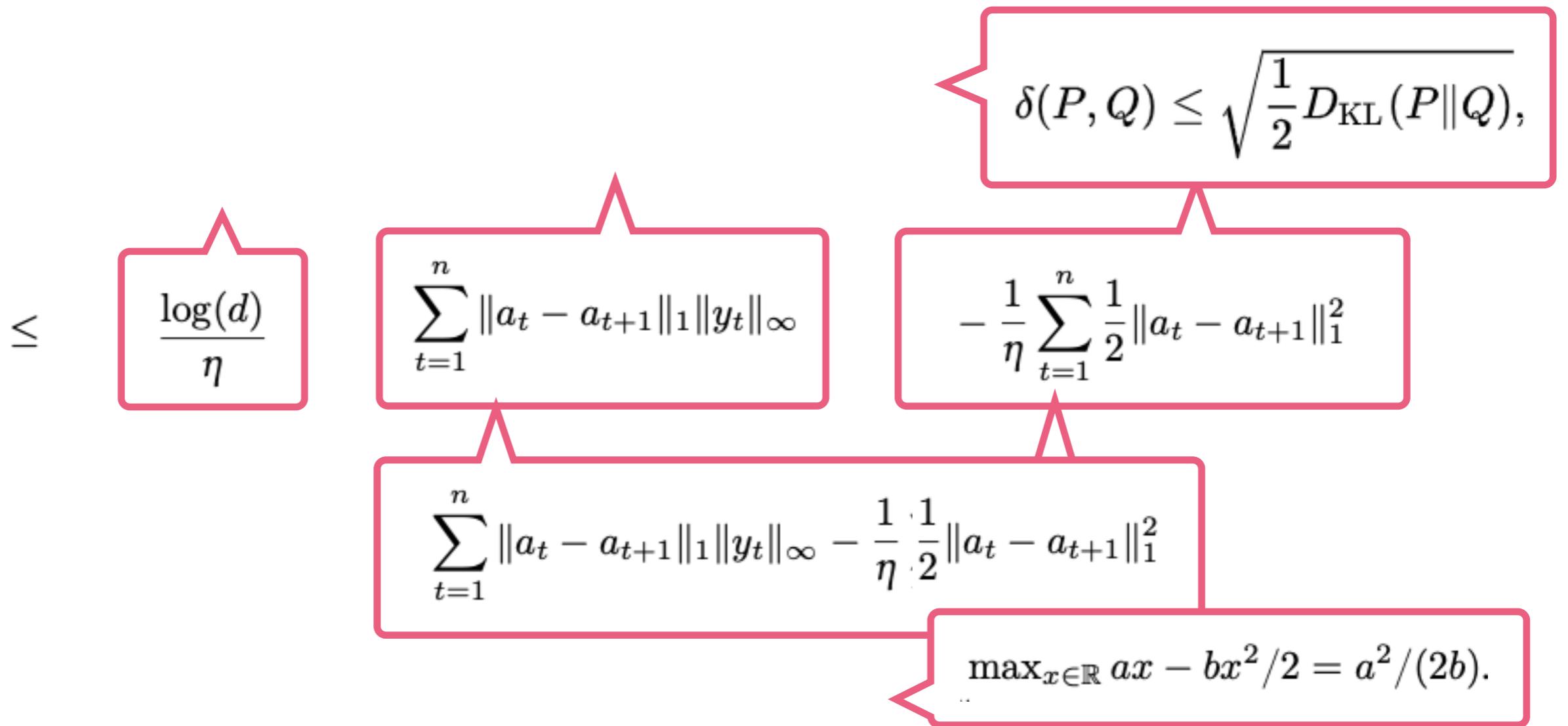
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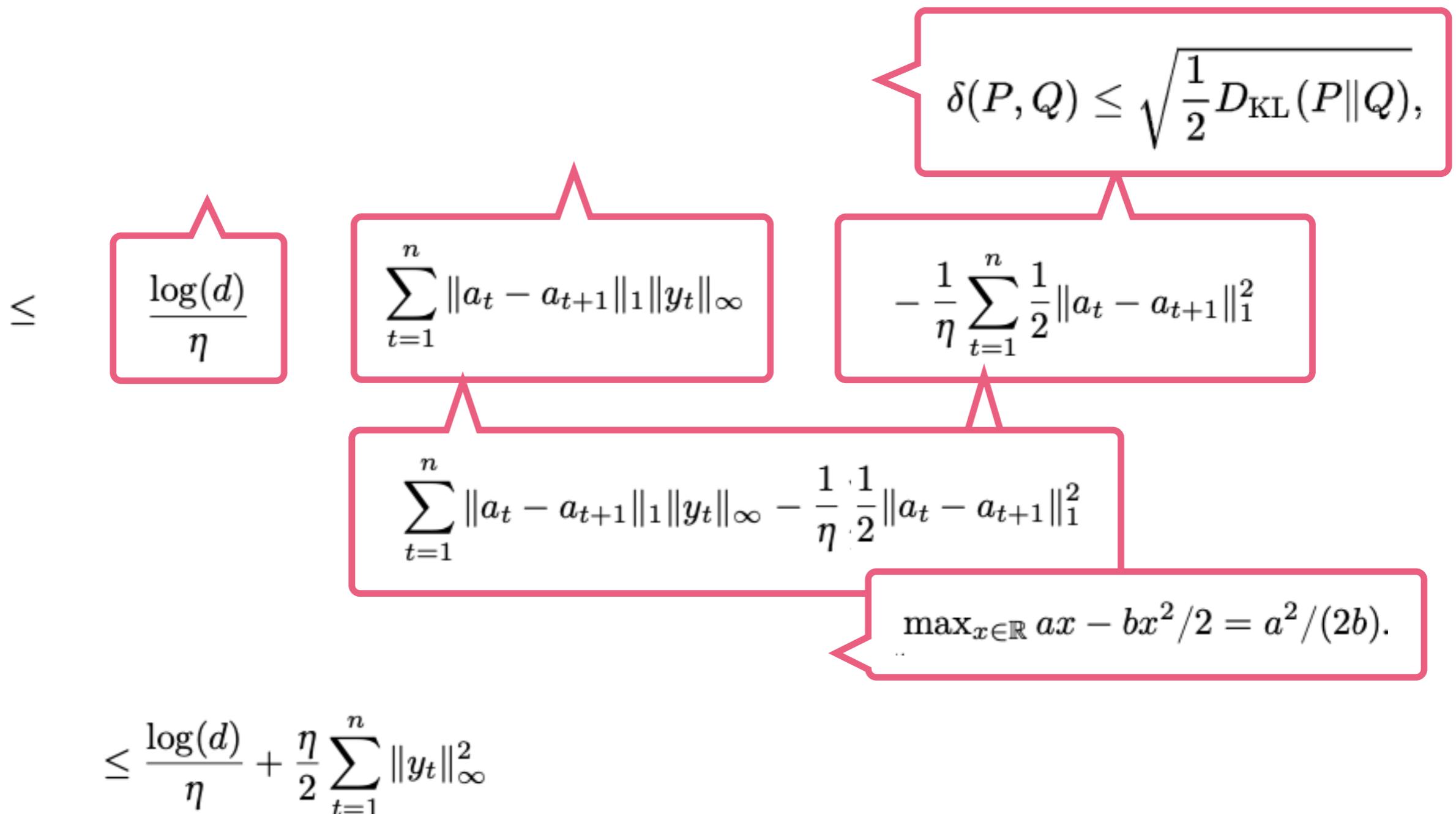
مثال کاهش آینه‌ای روی سادک

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$



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$$\delta(P, Q) \leq \sqrt{\frac{1}{2} D_{\text{KL}}(P \| Q)},$$

$$\leq \frac{\log(d)}{\eta} + \sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2$$

$$\sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \cdot \frac{1}{2} \|a_t - a_{t+1}\|_1^2$$

$$\max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).$$

$$\leq \frac{\log(d)}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_\infty^2 \leq \frac{\log(d)}{\eta} + \frac{\eta n}{2}$$

مثال کاهش آینه‌ای روی سادک

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$$\begin{aligned}
 & \leq \frac{\log(d)}{\eta} + \sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \sum_{t=1}^n \frac{1}{2} \|a_t - a_{t+1}\|_1^2 \\
 & \quad \leftarrow \frac{\log(d)}{\eta} \\
 & \quad \leftarrow \sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty \\
 & \quad \leftarrow \sum_{t=1}^n \|a_t - a_{t+1}\|_1 \|y_t\|_\infty - \frac{1}{\eta} \cdot \frac{1}{2} \|a_t - a_{t+1}\|_1^2 \\
 & \quad \leftarrow \max_{x \in \mathbb{R}} ax - bx^2/2 = a^2/(2b).
 \end{aligned}$$

$$\leq \frac{\log(d)}{\eta} + \frac{\eta}{2} \sum_{t=1}^n \|y_t\|_\infty^2 \leq \frac{\log(d)}{\eta} + \frac{\eta n}{2} = \sqrt{2n \log(d)}$$

نسخه پندپتی



بهینه‌سازی خطی برخط

$$\mathcal{L} \subset \mathbb{R}^d \quad \mathcal{A} \subset \mathbb{R}^d$$

تفاوت با یادگیری برخط

$$y_t \in L$$

- ۱ - انتخاب عمل $A_t \in A$
- ۲ - دریافت $\langle A_t, y_t \rangle$

$$R_n(a) = \mathbb{E} \left[\sum_{t=1}^n \langle A_t - a, y_t \rangle \right]$$

$$R_n = \max_{a \in \mathcal{A}} R_n(a)$$

چگونه تغییر بدھیم؟

کامپیوٹر
لرننگ

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

چگونه تغییر بدھیم؟

کامپیوٹر
لرننگ

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

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?
 y_t

چگونه تغییر بدھیم؟

کامپیوٹر
آرٹیفیشیونال

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} (\eta \langle a, y_t \rangle + D_F(a, a_t))$$

تصادف؟

? y_t

- 1: **Input** Legendre potential F , action set \mathcal{A} and learning rate $\eta > 0$
- 2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$
- 3: **for** $t = 1, \dots, n$ **do**
- 4: Choose measure P_t on \mathcal{A} with mean \bar{A}_t
- 5: Sample action A_t from P_t and observe $\langle A_t, y_t \rangle$
- 6: Compute estimate \hat{Y}_t of the loss vector y_t
- 7: Update:

$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \langle a, \hat{Y}_t \rangle + D_F(a, \bar{A}_t) \quad (\text{Mirror descent})$$

$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \sum_{s=1}^t \langle a, \hat{Y}_s \rangle + F(a) \quad (\text{follow-the-regularised-leader})$$

- 8: **end for**

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- 8: **end for**

تحلیل - از کران پشیمانی کاهش آینه‌ای

قضیه: پشیمانی کاهش آینه‌ای

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

تحلیل - از کران پشیمانی کاهش آینه‌ای

قضیه: پشیمانی کاهش آینه‌ای

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

امید؟

تحلیل - از کران پشیمانی کاهش آینه‌ای

قضیه: پشیمانی کاهش آینه‌ای

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

امید؟

? a_t

تحلیل - از کران پشیمانی کاهش آینه‌ای

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امید؟

? a_t

$$\mathbb{E}[E[\langle a_t, y_t \rangle]] = E[\langle \bar{A}_t, y_t \rangle]$$

تحليل

ناريبي

$$\mathbb{E} [\langle A_t, y_t \rangle] = \mathbb{E} [\langle \bar{A}_t, y_t \rangle] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, y_t \rangle | \bar{A}_t]] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, \hat{Y}_t \rangle | \bar{A}_t]]$$

تحليل

ناريبي

$$\mathbb{E} [\langle A_t, y_t \rangle] = \mathbb{E} [\langle \bar{A}_t, y_t \rangle] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, y_t \rangle | \bar{A}_t]] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, \hat{Y}_t \rangle | \bar{A}_t]]$$

$$R_n(a) = \mathbb{E} \left[\sum_{t=1}^n \langle A_t, y_t \rangle - \langle a, y_t \rangle \right] = \mathbb{E} \left[\sum_{t=1}^n \langle \bar{A}_t - a, \hat{Y}_t \rangle \right]$$

تحلیل

ناریبی

$$\mathbb{E} [\langle A_t, y_t \rangle] = \mathbb{E} [\langle \bar{A}_t, y_t \rangle] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, y_t \rangle | \bar{A}_t]] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, \hat{Y}_t \rangle | \bar{A}_t]]$$

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اگر بردارهای a_t داشته باشیم که با الگوریتم کاهش آینه‌ای براساس y_t تغییر کنند، داریم:

قضیه: پشیمانی کاهش آینه‌ای

$$R_n(a) \leq \frac{F(a) - F(a_1)}{\eta} + \sum_{t=1}^n \langle a_t - a_{t+1}, y_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(a_{t+1}, a_t).$$

تحلیل

ناریبی

$$\mathbb{E} [\langle A_t, y_t \rangle] = \mathbb{E} [\langle \bar{A}_t, y_t \rangle] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, y_t \rangle | \bar{A}_t]] = \mathbb{E} [\mathbb{E} [\langle \bar{A}_t, \hat{Y}_t \rangle | \bar{A}_t]]$$

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\tilde{Y}_t

\bar{A}_t

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تحلیل پشیمانی دو الگوریتم بندیت

$$R_n(a) \leq \mathbb{E} \left[\frac{F(a) - F(\bar{A}_1)}{\eta} + \sum_{t=1}^n \langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t) \right]$$

تحلیل پشیمانی دو الگوریتم بندیت

$$R_n(a) \leq \mathbb{E} \left[\frac{F(a) - F(\bar{A}_1)}{\eta} + \sum_{t=1}^n \langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t) \right]$$

$$R_n \leq \frac{\text{diam}_F(\mathcal{A})}{\eta} + \frac{1}{\eta} \sum_{t=1}^n \mathbb{E} \left[D(\bar{A}_t, \tilde{A}_{t+1}) \right]$$

$$\tilde{A}_{t+1} = \operatorname{argmin}_{a \in \text{dom}(F)} \eta \langle a, \hat{Y}_t \rangle + D_F(a, \bar{A}_t)$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$P_t = (1 - \gamma)\tilde{P}_t + \gamma\pi,$$

$$\tilde{P}_t(B) = \frac{\int_B \exp\left(-\eta \sum_{s=1}^{t-1} \hat{Y}_s(a)\right) da}{\int_{\mathcal{A}} \exp\left(-\eta \sum_{s=1}^{t-1} \hat{Y}_s(a)\right) da}.$$

مثال: بندیت خطی روی گوی واحد

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$$R_n \leq 2d\sqrt{3n(1 + \log_+(2n/d))}.$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

1: **Input** Legendre potential F , action set \mathcal{A} and learning rate $\eta > 0$

2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$

3: **for** $t = 1, \dots, n$ **do**

4: Choose measure P_t on \mathcal{A} with mean \bar{A}_t

5: Sample action A_t from P_t and observe $\langle A_t, y_t \rangle$

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$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \langle a, \hat{Y}_t \rangle + D_F(a, \bar{A}_t) \quad (\text{Mirror descent})$$

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8: **end for**

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$E_t \in \{0, 1\}$$

$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

$$U_t \sim \{\pm e_1, \dots, \pm e_d\}$$

$$A_t = E_t U_t + \frac{(1 - E_t) \bar{A}_t}{\|\bar{A}_t\|}$$

- 1: **Input** Legendre potential F , action set \mathcal{A} and learning rate η
- 2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$
- 3: **for** $t = 1, \dots, n$ **do**
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مثال: بندیت خطی روی گوی واحد

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1: **Input** Legendre potential F , action set \mathcal{A} and learning

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$$\hat{Y}_t = \frac{d E_t A_t \langle A_t, y_t \rangle}{1 - \|\bar{A}_t\|}$$

$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \sum_{s=1}^t \langle a, \hat{Y}_s \rangle + F(a) \quad (\text{follow-the-regularised-leader})$$

8: **end for**

$$E_t \in \{0,1\}$$

$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

$$\mathbb{E}_t[A_t] = \bar{A}_t$$

$$U_t \sim \{\pm e_1, \dots, \pm e_d\}$$

$$A_t = E_t U_t + \frac{(1-E_t)\bar{A}_t}{\|\bar{A}_t\|}$$

$$\hat{Y}_t = \frac{d\,E_t\,A_t\langle A_t,y_t\rangle}{1-\|\bar{A}_t\|}$$

$$E_t \,\in\, \{0,1\}$$

$$\mathbb{E}_t[E_t] \,=\, 1 - \|\bar{A}_t\|$$

$$\mathbb{E}_t[A_t] = \bar{A}_t$$

$$A_t = E_t U_t + \frac{(1-E_t)\bar{A}_t}{\|\bar{A}_t\|}$$

$$E[\hat{Y}_t]$$

$$\hat{Y}_t = \frac{d\,E_t\,A_t\langle A_t,y_t\rangle}{1-\|\bar{A}_t\|}$$

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$$E[\hat{Y}_t] = d \cdot E[A_t \langle A_t, y_t \rangle]$$

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$$A_t = E_t U_t + \frac{(1 - E_t) \bar{A}_t}{\|\bar{A}_t\|}$$

$$\begin{aligned} E[\hat{Y}_t] &= d \cdot E[A_t \langle A_t, y_t \rangle] \\ &= d \cdot \frac{1}{2d} \sum_i e_i \langle e_i, y_t \rangle - e_i \langle -e_i, y_t \rangle \end{aligned}$$

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$$E[\hat{Y}_t] = d \cdot E[A_t \langle A_t, y_t \rangle]$$

$$= d \cdot \frac{1}{2d} \sum_i e_i \langle e_i, y_t \rangle - e_i \langle -e_i, y_t \rangle$$

$$= d \cdot \frac{1}{2d} 2y_t$$

$$\hat{Y}_t = \frac{d E_t A_t \langle A_t, y_t \rangle}{1 - \|\bar{A}_t\|}$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\tilde{\mathcal{A}} = \{x \in \mathbb{R}^d : \|x\|_2 \leq r\}$$

1: **Input** Legendre potential F , action set \mathcal{A} and learning

2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$

3: **for** $t = 1, \dots, n$ **do**

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$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \sum_{s=1}^t \langle a, \hat{Y}_s \rangle + F(a) \quad (\text{follow-the-regularised-leader})$$

8: **end for**

$$E_t \in \{0, 1\}$$

$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

$$U_t \sim \{\pm e_1, \dots, \pm e_d\}$$

$$A_t = E_t U_t + \frac{(1 - E_t) \bar{A}_t}{\|\bar{A}_t\|}$$

$$\hat{Y}_t = \frac{d E_t A_t \langle A_t, y_t \rangle}{1 - \|\bar{A}_t\|}$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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1: **Input** Legendre potential F , action set \mathcal{A} and learning

2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$

3: **for** $t = 1, \dots, n$ **do**

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$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \langle a, \hat{Y}_t \rangle + D_F(a, \bar{A}_t)$$

$$E_t \in \{0, 1\}$$

$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

$$U_t \sim \{\pm e_1, \dots, \pm e_d\}$$

$$A_t = E_t U_t + \frac{(1 - E_t) \bar{A}_t}{\|\bar{A}_t\|}$$

$$\hat{Y}_t = \frac{d E_t A_t \langle A_t, y_t \rangle}{1 - \|\bar{A}_t\|}$$

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8: **end for**

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

$$\tilde{\mathcal{A}} = \{x \in \mathbb{R}^d : \|x\|_2 \leq r\}$$

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$$F(a) = -\log(1 - \|a\|) - \|a\|$$

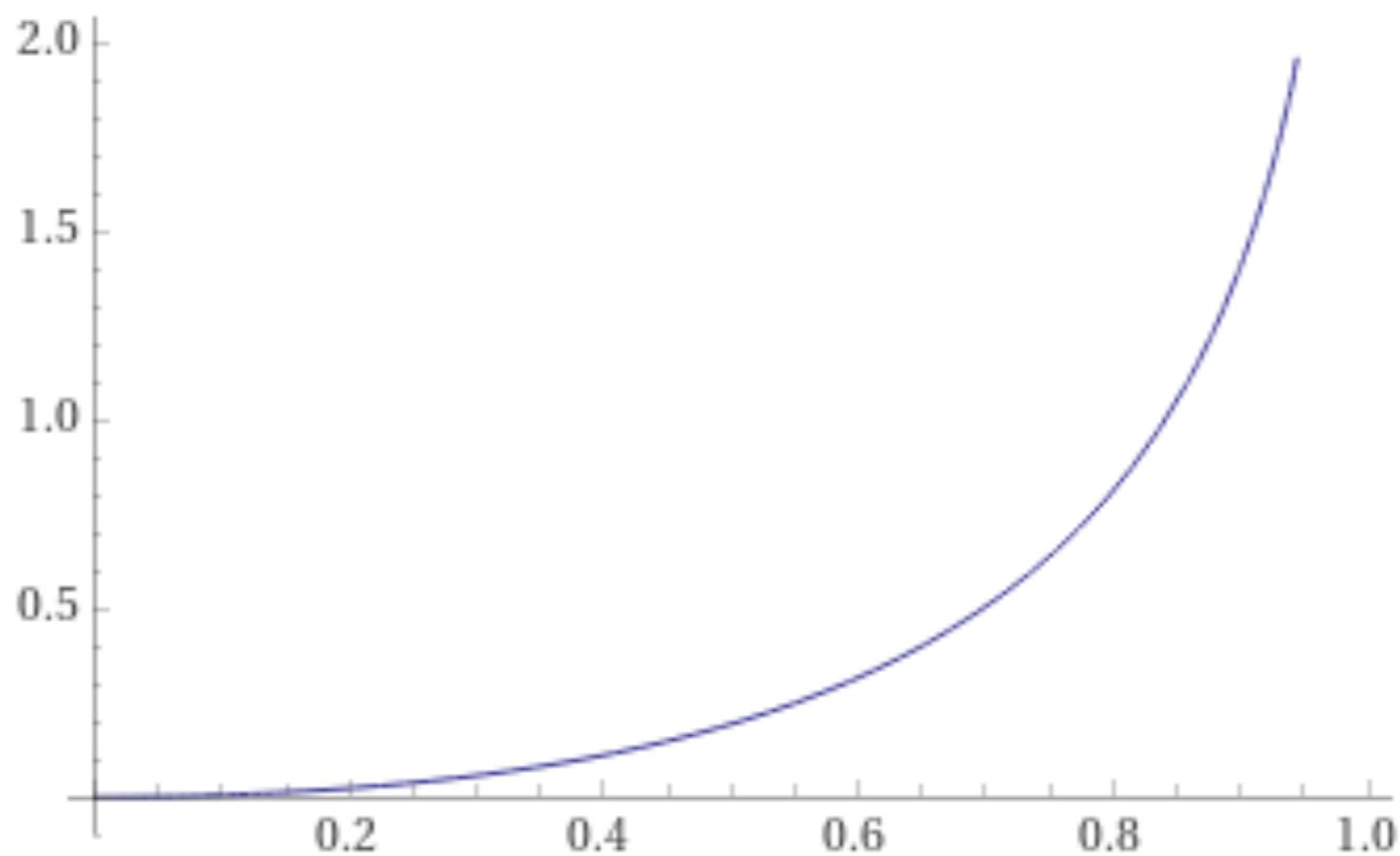
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$$F(a) = -\log(1 - \|a\|) - \|a\|$$

پیروی از پیش روی منظم شده بندیتی روی گوی واحد

پیش روی
از
منظم
شده

$$F(a) = -\log(1 - \|a\|) - \|a\|$$

$$a_1 = \operatorname{argmin}_{a \in \mathcal{A}} F(a)$$

$$a_{t+1} = \operatorname{argmin}_{a \in \mathcal{A}} \left(\eta \sum_{s=1}^t \langle a, y_s \rangle + F(a) \right)$$

$$\operatorname{arg min}_{x \in \tilde{\mathcal{A}}} \langle a, \eta \sum \hat{Y}_s \rangle - \log(1 - \|a\|) - \|a\|$$

$$\arg \min_{x \in \tilde{A}} \langle a, \eta \sum \hat{Y}_s \rangle - \log(1 - \|a\|) - \|a\|$$

شاید نقطه بهینه در قیود صدق کند:

$$\arg \min_{x \in \tilde{A}} \langle a, \eta \sum \hat{Y}_s \rangle - \log(1 - \|a\|) - \|a\|$$

شاید نقطه بهینه در قیود صدق کند:

$$\nabla(\dots) = \eta \sum \hat{Y}_s - \frac{1}{1 - \|a\|} \times \frac{-a}{\|a\|} - \frac{a}{\|a\|}$$

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$$\nabla(\dots) = 0$$

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موازی a
با اندازه‌ای مناسب

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- $\eta \sum \hat{Y}_s$ موازی a
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$$\frac{\|a\|}{1 - \|a\|} = \eta \left\| \sum \hat{Y}_s \right\|$$

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$$\frac{\|a\|}{1 - \|a\|} = \eta \left\| \sum \hat{Y}_s \right\| \rightarrow \frac{\|a\|}{1 - \|a\|} = \eta \left\| \sum \hat{Y}_s \right\| \rightarrow \frac{\|a\|}{1} = \frac{\eta \left\| \sum \hat{Y}_s \right\|}{1 + \eta \left\| \sum \hat{Y}_s \right\|}$$

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$$a = \frac{-\eta \hat{L}_{t-1}}{1 + \eta \left\| \hat{L}_{t-1} \right\|}$$

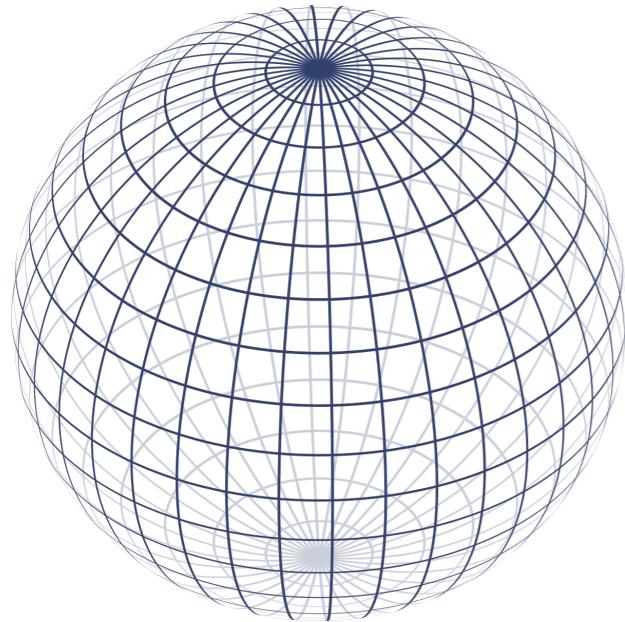
شاید نقطه بھینه در قیود صدق کند:

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شاید نقطه بهینه در قیود صدق کند:

اگر درون گوی نبود: بهینه: گرادیان | بردار

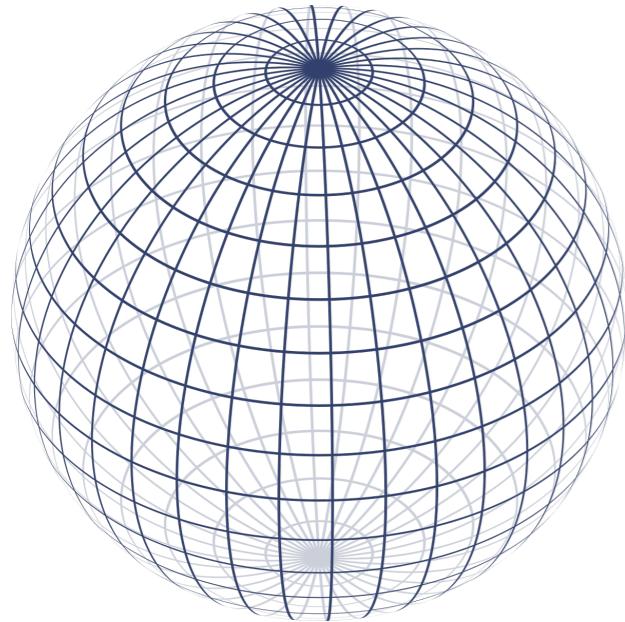
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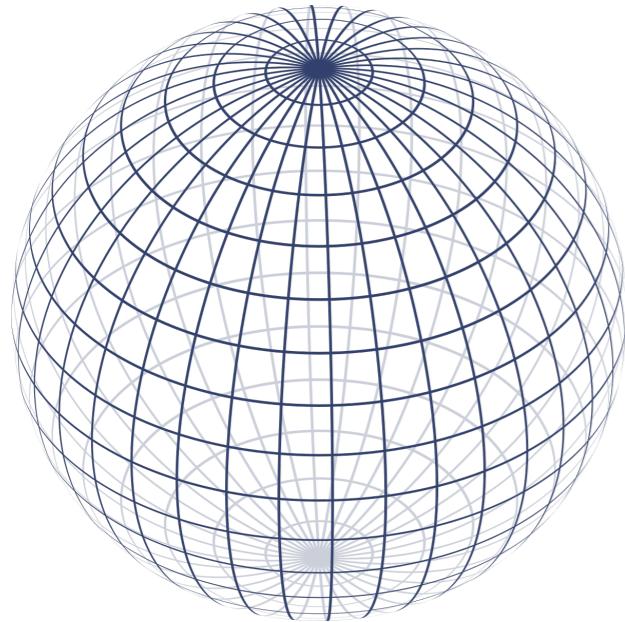


$$\eta \sum \hat{Y}_s + \frac{a}{1 - \|a\|} = \beta a$$

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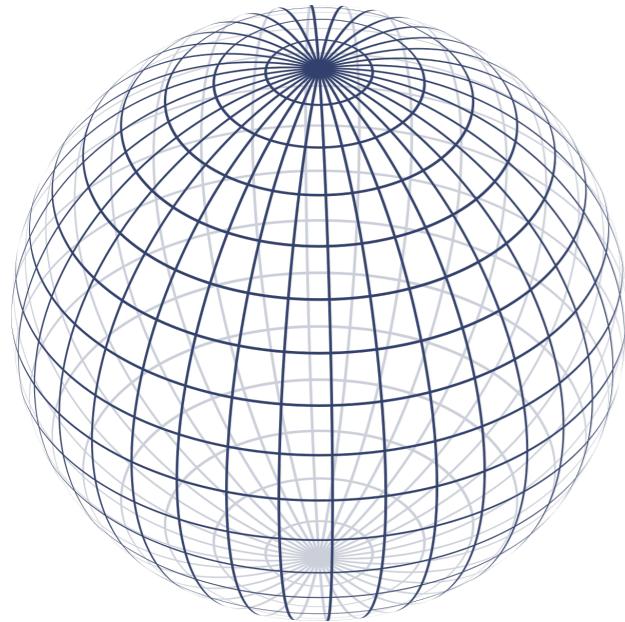


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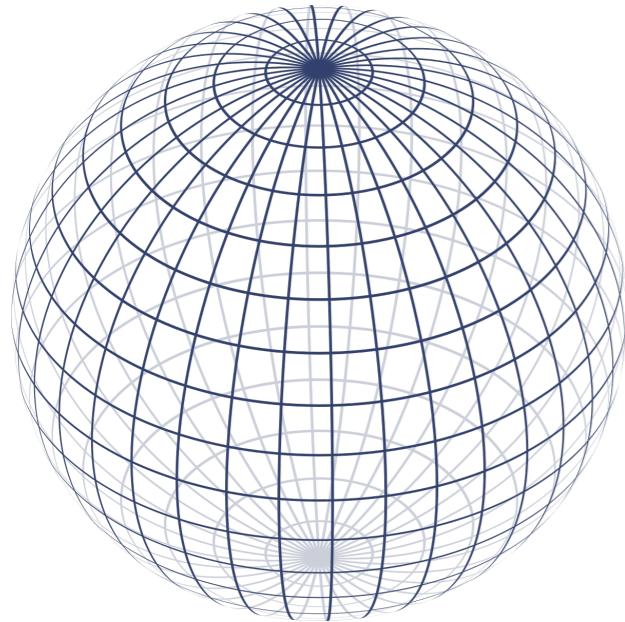
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موازی و روی سطح

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اگر درون گوی نبود: بهینه: گرادیان | بردار

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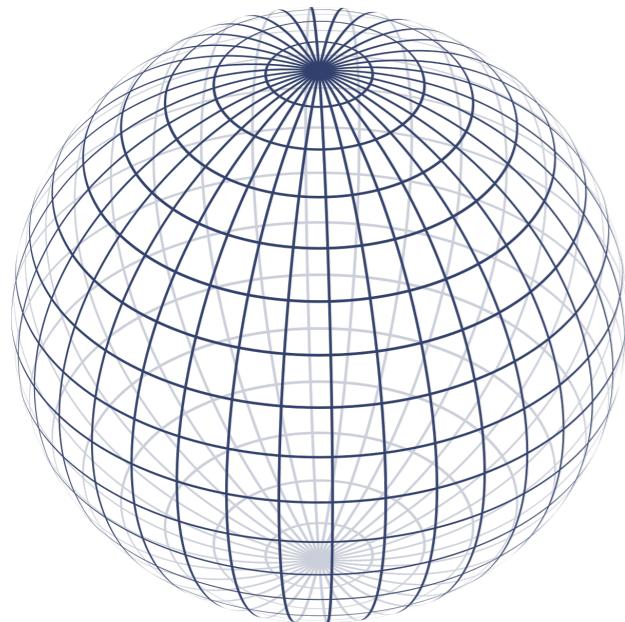
$$\eta \sum \hat{Y}_s = a(\beta - \frac{1}{1 - \|a\|})$$

موازی و روی سطح

تصویر روی کره

شاید نقطه بهینه در قیود صدق کند:

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$$\eta \sum \hat{Y}_s + \frac{a}{1 - \|a\|} = \beta a$$

$$\eta \sum \hat{Y}_s = a(\beta - \frac{1}{1 - \|a\|})$$

موازی و روی سطح

تصویر روی کره

$$\bar{A}_t = \Pi \left(\frac{-\eta \hat{L}_{t-1}}{1 + \eta \|\hat{L}_{t-1}\|} \right)$$

$$\hat{L}_{t-1} = \sum_{s=1}^{t-1} \hat{Y}_s,$$

مثال: بندیت خطی روی گوی واحد

$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\| \leq 1\}$$

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$$\bar{A}_{t+1} = \operatorname{argmin}_{\bar{A}_t} \bar{A}_t = \Pi \left(\frac{-\eta \hat{L}_{t-1}}{1 + \eta \|\hat{L}_{t-1}\|} \right), \bar{A}_t \right)$$

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$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

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(a) (follow-the-regularised-leader)

$$F(a) = -\log(1 - \|a\|) - \|a\|$$

تحلیل پشیمانی الگوریتم

$$R_n = \mathbb{E} \left[\sum_{t=1}^n \langle A_t - ra^*, y_t \rangle \right] + \sum_{t=1}^n \langle ra^* - a^*, y_t \rangle$$

تحلیل پشیمانی الگوریتم

$$R_n = \mathbb{E} \left[\sum_{t=1}^n \langle A_t - ra^*, y_t \rangle \right] + \sum_{t=1}^n \langle ra^* - a^*, y_t \rangle \leq R_n(ra^*) + (1-r)n$$

$$R_n(a) \leq \mathbb{E}\left[\frac{F(a)-F(\bar{A}_1)}{\eta} + \sum_{t=1}^n\langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t\rangle - \frac{1}{\eta}\sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t)\right]$$

$$R_n(a) \leq \mathbb{E} \left[\frac{F(a) - F(\bar{A}_1)}{\eta} + \sum_{t=1}^n \langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t) \right]$$

THEOREM 26.13. *Let $\eta > 0$ and f be Legendre and twice differentiable in $A = \text{int}(\text{dom}(f))$, $x, y \in A$, and let $z \in [x, y]$ be the point such that $D_f(x, y) = \frac{1}{2}\|x - y\|_{\nabla^2 f(z)}^2$. Then, for all $u \in \mathbb{R}^d$,*

$$\langle x - y, u \rangle - \frac{D_f(x, y)}{\eta} \leq \frac{\eta}{2} \|u\|_{(\nabla^2 f(z))^{-1}}^2.$$

$$R_n(a) \leq \mathbb{E} \left[\frac{F(a) - F(\bar{A}_1)}{\eta} + \sum_{t=1}^n \langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t) \right]$$

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$$R_n(ra^*) \leq \frac{\text{diam}_F(\tilde{\mathcal{A}})}{\eta} + \frac{\eta}{2} \mathbb{E} \left[\sum_{t=1}^n \|\hat{Y}_t\|_{(\nabla^2 F(Z_t))^{-1}}^2 \right]$$

$$R_n(a) \leq \mathbb{E} \left[\frac{F(a) - F(\bar{A}_1)}{\eta} + \sum_{t=1}^n \langle \bar{A}_t - \bar{A}_{t+1}, \hat{Y}_t \rangle - \frac{1}{\eta} \sum_{t=1}^n D(\bar{A}_{t+1}, \bar{A}_t) \right]$$

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$$\frac{1}{\eta} \log \left(\frac{1}{1-r} \right)$$

$$\mathbb{E}\left[\|\hat{Y}_t\|_{(\nabla^2 F(Z_t))^{-1}}^2\right] \leq$$

$$\mathbb{E}\left[\|\hat{Y}_t\|_{(\nabla^2 F(Z_t))^{-1}}^2\right] \leq~ E[\lambda_{\max} \|\hat{Y}_t\|^2]$$

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$$\mathbb{E}\left[\|\hat{Y}_t\|^2_{(\nabla^2F(Z_t))^{-1}}\right] \leq E[\lambda_{\max} \|\hat{Y}_t\|^2]$$

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$$\nabla^2 F(a) = \frac{I}{1 - \|a\|} + \frac{aa^t}{(1 - \|a\|)^2}$$

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اگر \bar{A}_{t+1} بیشینه نباشد، \bar{A}_{t+1} نزدیک‌تر است

برای \bar{A}_t ، تصویر رخ نداده

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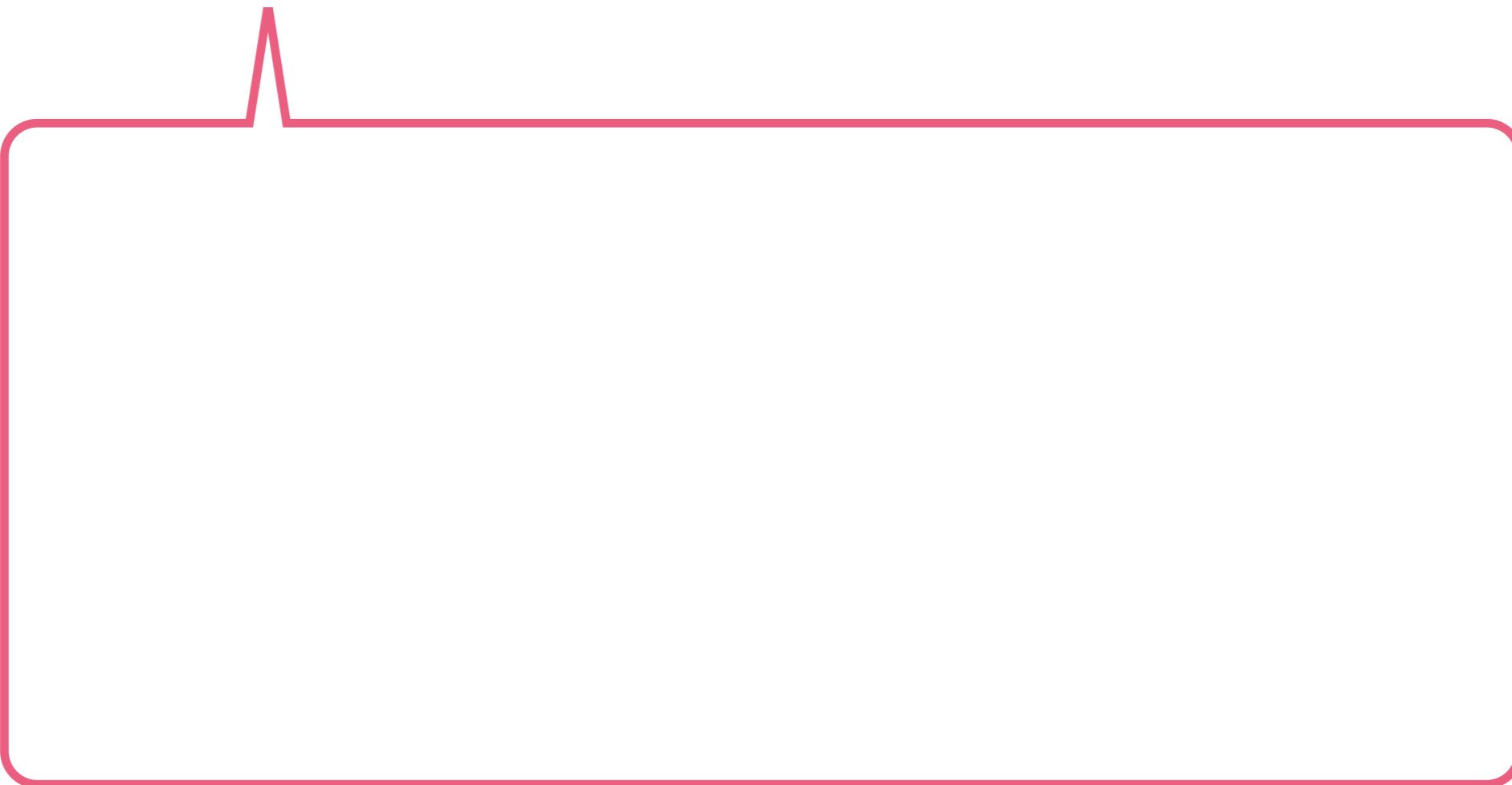
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$$R_n(ra^*) \leq \frac{\text{diam}_F(\tilde{\mathcal{A}})}{\eta} + \frac{\eta}{2} \mathbb{E} \left[\sum_{t=1}^n \|\hat{Y}_t\|_{(\nabla^2 F(Z_t))^{-1}}^2 \right]$$

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$$\eta = \sqrt{\log(n)/(3dn)}.$$

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مثال: بندیت خطی روی گوی واحد

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$$\mathcal{L} = \mathcal{A} = B_2^d = \{a \in \mathbb{R}^d : \|a\|_2 \leq 1\}$$

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$$\mathbb{E}_t[E_t] = 1 - \|\bar{A}_t\|$$

$$U_t \sim \{\pm e_1, \dots, \pm e_d\}$$

1: **Input** Legendre potential F , action set \mathcal{A} and learning

2: Choose $\bar{A}_1 = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} F(a)$

3: **for** $t = 1, \dots, n$ **do**

4: Choose measure P_t on \mathcal{A} with mean \bar{A}_t

5: Sample action A_t from P_t and observe $\langle A_t, y_t \rangle$

6: Compute estimate \hat{Y}_t of the loss vector y_t

7: Update:

$$\bar{A}_{t+1} = \operatorname{argmin}_{a \in \mathcal{A} \cap \operatorname{dom}(F)} \eta \langle a, \hat{Y}_t \rangle + D_F(a, \bar{A}_t)$$

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8: **end for**

$$F(a) = -\log(1 - \|a\|) - \|a\|$$

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بایان