

Лекция 4. Методы наращивания  
пространства градиента

$$\min_{x \in Q} f(x) \rightarrow x_* \text{ пер. зерн.}$$

$$Q = \mathbb{R}_+^d, Q = B_p^d(R), Q = S_d(1)$$

$$x^{k+1} = \pi_Q(x^k - h \nabla f(x^k))$$

$$\|x^{k+1} - x_*\|_2^2 = \|\pi_Q(x^k - h \nabla f(x^k)) - x_*\|_2^2 =$$

$$= \|\pi_Q(x^k - h \nabla f(x^k) - x_*)\|_2^2 \leq$$

$$\leq \|x^k - h \nabla f(x^k) - x_*\|_2^2 =$$

$$= \|x^k - x_*\|_2^2 - 2h \langle \nabla f(x^k), x^k - x_* \rangle + h^2 \|\nabla f(x^k)\|_2^2$$

$$\frac{1}{N} \sum_{k=0}^{N-1} \left\{ \underbrace{2h \langle \nabla f(x^k), x^k - x_* \rangle}_{\leq R} \leq \|x^k - x_*\|_2^2 - \|x^k - x_*\|_2^2 + h^2 \underbrace{\|\nabla f(x^k)\|_2^2}_{\leq M^2} \leq 2L(\bar{f}(x^k) - \bar{f}(x_*)) \right.$$

$$2h \frac{1}{N} \sum_{k=0}^{N-1} (\bar{f}(x^k) - \bar{f}(x_*)) \leq \overbrace{\|x^0 - x_*\|_2^2} - \|x^N - x_*\|_2^2 + \frac{1}{N} \sum_{k=0}^{N-1} h^2 \cdot \underbrace{\frac{M^2}{2L} (\bar{f}(x_k^N) - \bar{f}(x_*))}_{\leq \frac{MR^2}{N}}$$

$$\frac{1}{N} (2h - h^2) \sum_{k=0}^{N-1} (\bar{f}(x^k) - \bar{f}(x_*)) \leq R^2/N \rightarrow \frac{1}{N} \sum_{k=0}^{N-1} x^k$$

$$\frac{1}{N} (2h - h^2) (\bar{f}(\bar{x}^N) - \bar{f}(x_*)) \leq R^2/N \quad h = \frac{1}{2L}$$

$$\bar{f}(\bar{x}^N) - \bar{f}(x_*) \leq \frac{2LR^2}{N}; \quad \frac{MR}{\sqrt{N}}$$

$$f(\bar{x}^N) - f(x_*) \leq \frac{M_2 R_2}{\sqrt{N}}$$

$$\begin{aligned} \frac{1}{p} + \frac{1}{q} &= 1 \\ M_p &= \max_{x \in Q} \|Df(x)\|_2 \\ M_1 &= \max_{x \in Q} \|Df(x)\|_\infty \end{aligned}$$

$M_2 \rightarrow |f(y) - f(x)| \leq M_2 \|y - x\|_2$   
 $R_2 \rightarrow \|x^0 - x_*\|_2$   
 $M_1 \leq M_2$   
 $R_1 \rightarrow \|x^0 - x_*\|_1 \leq R_2$

$$M_2 = \max_{x \in Q} \|Df(x)\|_2$$

$$M_1 \sqrt{d} \approx M_2 \quad \text{max}_{x \in Q} \|x\|_2$$

$$\begin{aligned} \|\underbrace{(1, \dots, 1)}_d\|_\infty &= 1 \\ \|\underbrace{(1, \dots, 1)}_d\|_2 &= \sqrt{d} \end{aligned}$$

$$\|Df(y) - Df(x)\|_2 \leq$$

$$\leq L \|y - x\|_p$$

$$\frac{1}{p} + \frac{1}{q} = 1$$

$$p=q=2$$

$$p=1, q=\infty$$

$$L_p = \max_{x \in Q} \max_{\|h\|_p \leq 1} \langle h, Df(x)h \rangle$$

$$\boxed{\frac{L_2}{\sqrt{d}} \leq L_1 \leq L_2}$$

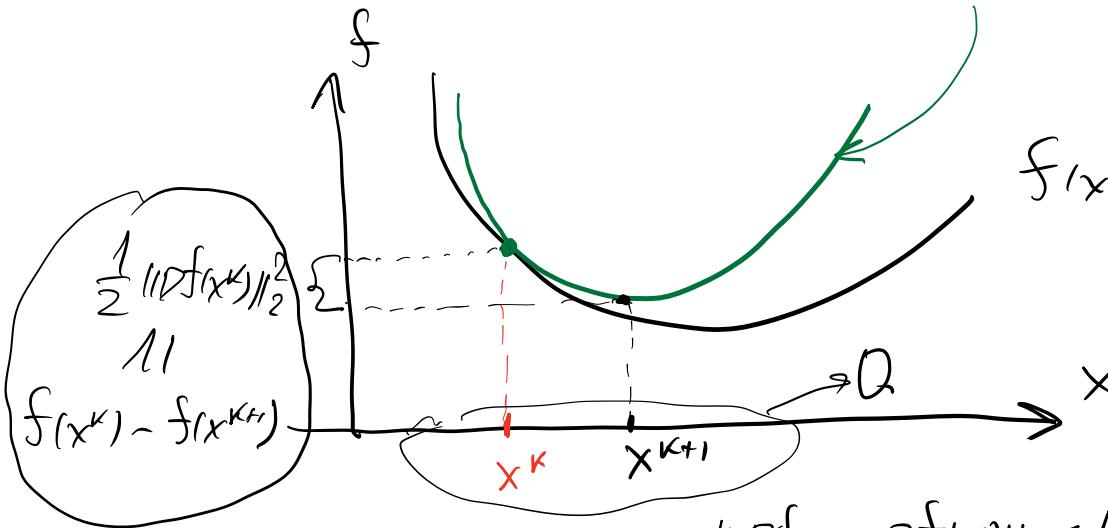
$$\boxed{\frac{M_2}{\sqrt{d}} \leq M_1 \leq M_2}$$

$$f(\hat{x}^N) - f(x_*) \leq \frac{4R_p^2}{N^\alpha} \quad d=1, 2$$

$\nearrow \text{newer}$   
 $\searrow \text{year}$

$$\frac{M_2 R_p}{\sqrt{N}}$$

$$x^{k+1} = \underset{x \in Q}{\operatorname{argmin}} \left\{ f(x^k) + \langle \nabla f(x^k), x - x^k \rangle + \frac{\mu}{2} \|x - x^k\|_2^2 \right\} = \Pi_Q \left( x^k - \frac{1}{\mu} \nabla f(x^k) \right)$$



$$\|Df(y) - Df(x)\|_2 \leq L \|y - x\|_2$$

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{L}{2} \|y - x\|_2^2$$

$$\frac{1}{2L} \|Df(x^k)\|_2^2 \leq f(x^k) - f(x^{k+1})$$

$$f(x^n) - f(x_*) \leq \frac{L \rho^2}{N}$$

A.C. Heringspunt  
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$$\frac{1}{2} \|x - x^k\|_2^2 = \frac{1}{2} \|x\|_2^2 - \frac{1}{2} \|x^k\|_2^2 - \langle \nabla \left( \frac{1}{2} \|x\|_2^2 \right) \Big|_{x=x^k}, x - x^k \rangle$$

$\frac{1}{2} \|x\|_2^2 \rightarrow d(x) - 1$ -eindige form. omw. p-hyp Mtr  
ha Q  $p \in [1, 2]$

$$V(x, x^*) = f(x) - f(x^*) - \langle \nabla f(x^*), x - x^* \rangle$$

→ универсальная формула

1-норма,  $Q = S_d(\mathbb{Z})$

$$f(x) = \sum_{i=1}^d x_i \ln x_i - \text{экспоненциальная}$$

→ 1-норма есть о-норма для

1-нормы на  $S_d(\mathbb{Z})$

$$\mu_p = \max_{x \in Q} \frac{m/n}{\|x\|_p \leq 1} \underbrace{\langle h, \nabla f(x) h \rangle}_{\begin{bmatrix} x_1 & \dots & x_n \end{bmatrix}} \quad p=1$$

$$KL(x, x^*) = \sum_{k=1}^d x_k \ln(x_k/x_k^*)$$

$$x^{k+1} = \underset{x \in S_d(\mathbb{Z})}{\operatorname{argmin}} \left\{ f(x^*) + \langle \nabla f(x^*), x - x^* \rangle + \underbrace{L_p V(x, x^*)}_{{KL}(x, x^*)} \right\} \rightarrow$$

$$f(y) \leq f(x) + \langle \nabla f(x), y - x \rangle + \frac{KL(x, x^*)}{2} \|y - x\|_p^2 \leq$$

$$\leq f(x) + \langle \nabla f(x), y - x \rangle + L_p \|y - x\|_p \quad \|\nabla f(y) - \nabla f(x)\|_p \leq L_p \|y - x\|_p$$

$$+ L_p \|y - x\|_p \quad \frac{1}{p} + \frac{1}{q} = 1$$

$$\frac{V}{\frac{1}{p} \|y - x\|_p^2} \rightarrow x_i^{k+1} = \frac{\exp(-\frac{\partial f(x^*)/\partial x_i}{L})}{\sum_{j=1}^d \exp(-\frac{\partial f(x^*)/\partial x_j}{L})}$$

$$O(t)$$

$$f(\bar{x}^n) - f(x_*) \leq \frac{M_p \bar{R}_p}{\sqrt{N}} + \underbrace{\frac{L_p \bar{R}_p^2}{N^\alpha}, \alpha=1,2}_{\bar{R}_p^2 = 2V(x_*, x^*)}$$

$$\bar{R}_p^2 = \|x^0 - x_*\|_p^2$$

Defn:  $\|x^0 - x_*\|_p^2 \geq V(x_*, x^*) \geq \frac{1}{2} \|x^0 - x_*\|_p^2$

$$Q = S_d(1)$$

$$\max_{x_*} KL(x^0, x_*) \leq 2\ln d$$

$\left( \begin{array}{c} 1 \\ \vdots \\ n \end{array} \right)$   
 ↴ versusp  
customers

$$\max_{x^0, x_* \in S_d(1)} \|x^0 - x_*\|_1^2 \leq 2$$

$\left( \begin{array}{c} 1 \\ 0 \\ \vdots \\ 0 \end{array} \right) \quad \left( \begin{array}{c} 0 \\ 1 \\ \vdots \\ 0 \end{array} \right)$

$$\| \left[ \begin{array}{c} x \\ x \\ \vdots \\ x \end{array} \right] 1 \|_1^2 = 1$$

$$\| \left[ \begin{array}{c} 0 \\ 0 \\ \vdots \\ 0 \end{array} \right] 1 \|_2^2 = 1$$

$$x^{k+1} = \text{Mirr}_{x^k}^Q (\underline{\text{hdf}(x^k)}) =$$

$$= \underset{x \in Q}{\arg \min} \left\{ f(x^k) + h \langle \nabla f(x^k), x - x^k \rangle + V(x, x^k) \right\}$$

$$\frac{1}{p} + \frac{1}{2} = 1$$

$\nabla f(x) - \text{curves from } p\text{-norm}$

$$V(x^{k+1}, x_*) \leq V(x^k, x_*) - 2h \langle \nabla f(x^k), x^k - x_* \rangle + h^2 \|Df(x^k)\|_2^2$$

$$\frac{1}{2} \|x^{k+1} - x_*\|_2^2 \leq \frac{1}{2} \|x^k - x_*\|_2^2 - 2h \langle \nabla f(x^k), x^k - x_* \rangle + h^2 \|Df(x^k)\|_2^2$$

am. §3 nosovne MChMDO.

Пример (Линейное).

$$\min_{x \in Q} f(x) \rightarrow \frac{1}{2} \langle x | Ax \rangle - \langle b | x \rangle$$

$$Q = \{x \in \mathbb{R}^d : \sum_i x_i = 1\}$$

$$x^{k+1} = \underbrace{\Pi_Q(x^k - h \nabla f(x^k))}_{\begin{array}{l} \xrightarrow{\text{Л}} \\ \text{Ax} - b \\ O(d^2) \end{array}}$$

$$f(\bar{x}^n) - f(x_*)$$

$$L_2 = \max_{x \in Q} \max_{h: \|h\|_2 \leq 1} \underbrace{\langle h, Df(x)h \rangle}_A = \lambda_{\max}(A)$$

$$R^2 = \|x^0 - x_*\|_2^2 \leq 2$$

$$\underbrace{O(d^2 + d)}_{\text{constant. неравн}} \quad f(\bar{x}^n) - f(x_*) \leq \frac{(LR^2)}{N} \underbrace{\frac{2\lambda_{\max}(A)}{N}}$$

$p=2$

$$N = \frac{\lambda_{\max}(A) \cdot 2}{\varepsilon}, \quad O(\delta^2)$$

$p=1$

$$N = \frac{\max|A_{ij}| \cdot 2 \ln \delta}{\varepsilon}, \quad O(\delta^2)$$

$$L_1 = \max_{h: \|h\|_1 \leq 1} \langle h, Ah \rangle = \max_{i,j} |A_{ij}| \leq \lambda_{\max}(A)$$

$$R^2 \leq 2 \ln \delta$$

$$M_p R_p \quad 1 \leq p \leq 2$$

$$\|Df(x)\|_q \rightarrow \|x^0 - x_*\|_p \text{ lnd}$$

$$\|Df(x)\|_\infty \leq \|Df(x)\|_2 \quad p=1 \\ \|x^0 - x_*\|_1 \geq \|x^0 - x_*\|_2$$

### Универсальная мем

Мем магнитных трансформаторов

$$y^0 = z^0, \quad A_0 = \alpha_0 = 0, \quad \alpha_{K+1} = \frac{1}{2L} \sqrt{\frac{1}{4L^2} + \alpha_K^2}$$

$$A_{K+1} = A_K + \alpha_{K+1}$$

$$\min_{x \in Q} f(x)$$

$$x^{k+1} = \frac{\alpha_{K+1} z^K + A_K y^K}{A_{K+1}}$$

$$z^{k+1} = \underset{x \in Q}{\operatorname{argmin}} \left\{ \alpha_{K+1} \langle \nabla f(x^{k+1}), x - x^{k+1} \rangle + V(x, z^K) \right\}$$

$$y^{K+1} = \frac{\alpha_{K+1} z^{K+1} + A_K y^K}{A_{K+1}}$$

$$f(y^*) - f(y_*) \leq \frac{4LR^2}{N^2}$$

$$R^2 = V(x, x^*) \quad \begin{array}{l} \xrightarrow{\text{норма}} f(x) \\ \downarrow \begin{array}{l} \text{1-момент} \\ \text{без} \\ \partial x. \\ p-\text{норма} \end{array} \end{array}$$

$$\|Df(y) - Df(x)\|_p \leq L \|y - x\|_p$$

$$\min_{x \in Q} f(x)$$

$$\cancel{Df(x)} \rightarrow Df(x, \xi) \quad [D_x f(x, \xi)]$$

$$E[Df(x, \xi)] = Df(x) \quad \forall x \in Q$$

$$E[\|Df(x, \xi)\|_p^2] \leq M^2 \quad \forall x \in Q$$

$$p \in [1, 2] \quad \frac{1}{p} + \frac{1}{2} = 1$$

$$h = \frac{\epsilon}{M^2}$$

$$x^{k+1} = \text{Merr}_{x^k}^Q(h \nabla f(x^k, z^k))$$

$$\boxed{V(x_*, x^{k+1}) \leq V(x_*, x^k) - 2h \langle \zeta, x^k - x_* \rangle + h^2 \| \zeta \|^2_M}$$

$$2h \langle \nabla f(x^k, z^k), x^k - x_* \rangle \leq V(x_*, x^k) - V(x_*, x^{k+1}) + h^2 \|\nabla f(x^k, z^k)\|_2^2 \quad \mathbb{E}_{\zeta^k} \sum_i \|x_i^k\|$$

$$x^k(z^0, \dots, z^{k-1}) \quad x^k \text{ не является } \sigma z^k$$

$$2h \underbrace{\langle \nabla f(x^k), x^k - x_* \rangle}_{f(x^k) - f(x_*)} \leq V(x_*, x^k) - V(x_*, x^{k+1}) + h^2 M^2 \quad \left. \begin{array}{l} \\ \end{array} \right\} \mathbb{E}_{x^k}$$

$$\mathbb{E}\left[2h(f(x^k) - f(x_*))\right] \leq \mathbb{E}V(x_*, x^k) - \mathbb{E}V(x_*, x^{k+1}) + h^2 M^2 \quad \sum_{k=0}^m$$

$$2h N \left[ \mathbb{E}[f(\bar{x}^m)] - f(x_*) \right] \leq V(x_*, x^0) + h^2 M^2 N$$

$$|E[f(\bar{x}^n)] - f(x_*)| \leq \underbrace{\frac{R^2}{2hn}}_{h = \frac{R}{M\sqrt{n}}} + \frac{hM^2}{2} = \frac{MR}{\sqrt{n}}$$

$\approx \frac{\varepsilon}{M^2}$