

$$\min_{x \in \mathbb{R}^d} f(x)$$

$$f(x) = \frac{1}{2} x^T A x - b^T x$$

\uparrow
 симм.
 $\mathbb{R}^{d \times d}$

$$\nabla f(x) = \underbrace{Ax - b}_{\substack{d^2 \text{ операций} \\ + d \text{ операций}}} = \underbrace{\quad}_{O(d^2)}$$

$$f(x) = \frac{1}{2n} \sum_{i=1}^n (a_i^T x - b_i)^2 \leftarrow \text{линейное регрессия}$$

SGD (вектор системы)

$$\begin{aligned} \nabla f_i(x) &= \nabla \left(\frac{1}{2} (a_i^T x - b_i)^2 \right) \\ &= (a_i^T x - b_i) a_i \end{aligned}$$

вычисляем $\nabla f_i(x)$ генерируя
в n раз по сравнению
с $\nabla f(x)$

Координатный спуск

$$\begin{aligned} \nabla f_{(j)}(x) &= \left(\frac{1}{2n} \sum_{i=1}^n (a_{i1} x_1 + a_{ij} x_j + \dots - b_i)^2 \right)_j \\ &= \frac{1}{n} \sum_{i=1}^n (a_i^T x - b_i) a_{ij} \end{aligned}$$

вычисляем / производим $\nabla f_{(j)}(x)$
генерируя в d раз по сравнению
с $\nabla f(x)$

SGD (сумм.)

$$x^{(k+1)} = x^{(k)} - \gamma \nabla f_{i_k}(x^{(k)})$$

\uparrow
 случайный
выбор i_k
из 1 до n

Координатный спуск

$$x^{(k+1)} = x^{(k)} - \gamma \nabla f_{(j_k)}(x^{(k)})$$

\uparrow
 случайный
выбор коорд.
из 1 до d

$\begin{pmatrix} 0 \\ \vdots \\ 0 \\ f_{(j_k)}'(x^{(k)}) \\ 0 \\ \vdots \\ 0 \end{pmatrix}$

Достоинства координатного

Преимущества:

- f - μ -сильно выпуклая
- f - L -ограничена

Dox-ba:

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &= \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - 2\gamma \mathbb{E}[\langle \nabla f_{(j_k)}(x^k); x^k - x^* \rangle] \\ &\quad + \gamma^2 \mathbb{E}[\|\nabla f_{(j_k)}(x^k)\|_2^2] \quad (*) \end{aligned}$$

$\mathbb{E}[\langle \nabla f_{(j_k)}(x^k); x^k - x^* \rangle]$ + tower property:

$$\mathbb{E}[\mathbb{E}[\langle \nabla f_{(j_k)}(x^k); x^k - x^* \rangle | x^k]] \ominus$$

$$\begin{aligned} \mathbb{E}[\nabla f_{(j_k)}(x^k) | x^k] &= \sum_{i=1}^d \mathbb{P}\{j_k = i\} \nabla f_{(i)}(x^k) \\ &= \frac{1}{d} \sum_{i=1}^d \nabla f_{(i)}(x^k) \\ &\quad \underbrace{\qquad\qquad\qquad}_{\nabla f(x^k)} \\ &= \frac{1}{d} \nabla f(x^k) \end{aligned}$$

$$\ominus \mathbb{E}\left[\frac{1}{d} \langle \nabla f(x^k); x^k - x^* \rangle\right] \quad (**)$$

Lemma $(**)$ b $(*)$:

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &= \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - \frac{2\gamma}{d} \mathbb{E}[\langle \nabla f(x^k); x^k - x^* \rangle] \\ &\quad + \gamma^2 \mathbb{E}[\|\nabla f_{(j_k)}(x^k)\|_2^2] \quad (***) \end{aligned}$$

$\mathbb{E}[\|\nabla f_{(j_k)}(x^k)\|_2^2]$ + tower property

$$\mathbb{E}[\mathbb{E}[\|\nabla f_{(j_k)}(x^k)\|_2^2 | x^k]] =$$

$$= \mathbb{E} \left[\sum_{i=1}^d \frac{1}{d} \|\nabla f_{(i)}(x^k)\|_2^2 \right]$$

$$= \mathbb{E} \left[\frac{1}{d} \sum_{i=1}^d \|\nabla f_{(i)}(x^k)\|_2^2 \right]$$

$$\sum_{i=1}^d (f'_{(i)}(x^k))^2 = \|\nabla f(x^k)\|_2^2$$

$$= \mathbb{E} \left[\frac{1}{d} \|\nabla f(x^k)\|_2^2 \right] \quad (****)$$

Тезисы (****) и (****)

$$\begin{aligned} \mathbb{E} [\|x^{k+1} - x^*\|_2^2] &= \mathbb{E} [\|x^k - x^*\|_2^2] \\ &\quad - \frac{2\gamma}{d} \mathbb{E} [\langle \nabla f(x^k); x^k - x^* \rangle] \\ &\quad + \gamma^2 \mathbb{E} [\|\nabla f(x^k)\|_2^2] \end{aligned}$$

L-функция, μ-функция

$$\begin{aligned} \mathbb{E} [\|x^{k+1} - x^*\|_2^2] &\leq \mathbb{E} [\|x^k - x^*\|_2^2] \\ &\quad - \frac{2\gamma}{d} \mathbb{E} \left[\frac{\mu}{2} \|x^k - x^*\|_2^2 + f(x^k) - f(x^*) \right] \\ &\quad + \gamma^2 \mathbb{E} [2L (f(x^k) - f(x^*))] \\ &= (1 - \frac{\gamma\mu}{d}) \mathbb{E} [\|x^k - x^*\|_2^2] \\ &\quad - \frac{2\gamma}{d} (1 - L\gamma) \underbrace{\mathbb{E} [f(x^k) - f(x^*)]}_{\geq 0} \end{aligned}$$

$$\gamma \leq \frac{1}{L}$$

$$\mathbb{E} [\|x^{k+1} - x^*\|_2^2] \leq (1 - \frac{\gamma\mu}{d}) \mathbb{E} [\|x^k - x^*\|_2^2]$$

$\gamma = \frac{1}{L}$ и заданные параметры

Среднее робастное

$$\mathbb{E}[\|x^{(k+1)} - x^*\|_2^2] \leq \left(1 - \frac{\mu}{dL}\right)^{k+1} \mathbb{E}[\|x^0 - x^*\|_2^2]$$

где μ — минимальное значение ϵ функции

$$O\left(\frac{dL}{\mu} \cdot \log \frac{\|x^0 - x^*\|_2^2}{\epsilon}\right) \text{ итераций}$$

⊕ итерация генерирует d раз (где d — размер задачи), чем $y \in D$

⊖ раз-то итерация работает d раз

$$\min_{x \in \mathbb{R}^d} x_1^2 + x_2^2 + \dots + x_d^2$$

SAGA

$$y_i^k = \begin{cases} \nabla f_{i_k}(x^k) & i = i_k \\ y_i^{k-1} & i \neq i_k \end{cases}$$

$$1) x^{(k+1)} = x^k - \gamma \cdot \frac{1}{n} \sum_{i=1}^n y_i^k$$

$$2) x^{(k+1)} = x^k - \gamma \cdot \left(\frac{1}{n} \sum_{i=1}^n y_i^{k-1} + \nabla f_{i_k}(x^k) - y_{i_k}^{k-1} \right)$$

SAGA

модель модели макс

SEGA h^k — берем на шаг

$$h_{(i)}^k = \begin{cases} \langle \nabla f(x^k); e_{i_k} \rangle & i = i_k \\ h_{(i)}^{k-1} & i \neq i_k \end{cases}$$

$$1) x^{(k+1)} = x^k - \gamma h^k$$

$$2) x^{(k+1)} = x^k - \gamma \cdot (h^{k-1} + d(\nabla f_{i_k}(x^k) - h_{(i_k)}^{k-1}))$$

SEGA

Док. об эффективности:

Предположим

- f — μ -модель функции
- f — L -выпукло

Donc-bo: $x^{k+1} = x^k - \gamma g^k$

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &= \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - 2\gamma \mathbb{E}[\langle g^k; x^k - x^* \rangle] \\ &\quad + \gamma^2 \mathbb{E}[\|g^k\|_2^2] \end{aligned} \quad (*)$$

$\mathbb{E}[\langle g^k; x^k - x^* \rangle]$ + tower property

$$\mathbb{E}[\mathbb{E}[\langle g^k; x^k - x^* \rangle | x^k]] \stackrel{(*)}{=}$$

$$\mathbb{E}[g^k | x^k] \stackrel{(*)}{=} 1) \quad g^k = h^k$$

$$= \mathbb{E}[h^k | x^k] = \begin{pmatrix} \mathbb{E}[h_{(i)}^k | x^k] \\ \vdots \end{pmatrix} = \begin{pmatrix} \frac{1}{d} \langle \nabla f(x^k); e_i \rangle + (1 - \frac{1}{d}) h_{(i)}^{k-1} \\ \vdots \end{pmatrix}$$

$$= \frac{1}{d} \underbrace{\begin{pmatrix} \langle \nabla f(x^k); e_i \rangle \\ \vdots \end{pmatrix}}_{\nabla f(x^k)} + (1 - \frac{1}{d}) \underbrace{\begin{pmatrix} h_{(i)}^{k-1} \\ \vdots \end{pmatrix}}_{h^{k-1}}$$

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

$$\stackrel{(*)}{=} 2) \quad g^k = h^{k-1} + d(\nabla f_{(i_k)}(x^k) - h_{(i_k)}^{k-1})$$

$$= \mathbb{E}[h^{k-1} + d(\nabla f_{(i_k)}(x^k) - h_{(i_k)}^{k-1}) | x^k]$$

$$= h^{k-1} + d \mathbb{E}[\nabla f_{(i_k)}(x^k) - h_{(i_k)}^{k-1} | x^k]$$

$$= h^{k-1} + d \cdot \frac{1}{d} (\nabla f(x^k) - h^{k-1}) = \boxed{\nabla f(x^k)} \quad (**)$$

Тезисы (**) & (*):

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &= \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - 2\gamma \mathbb{E}[\langle \nabla f(x^k); x^k - x^* \rangle] \\ &\quad + \gamma^2 \mathbb{E}[\|g^k\|_2^2] \end{aligned}$$

μ -curvature bounds

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &= \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - 2\gamma \mathbb{E}\left[\frac{\mu}{2} \|x^k - x^*\|_2^2 + f(x^k) - f(x^*)\right] \\ &\quad + \gamma^2 \mathbb{E}[\|g^k\|_2^2] \quad (**) \end{aligned}$$

$\mathbb{E}[\|g^k\|_2^2]$ + tower property

$$\begin{aligned} \mathbb{E}[\|g^k\|_2^2 | x^k] &= \mathbb{E}[\|g^k - \nabla f(x^*)\|_2^2 | x^k] \\ &= \mathbb{E}[\|h^{k-1} + d(\nabla f_{(i_k)}(x^k) - h_{(i_k)}^{k-1}) - \nabla f(x^*)\|_2^2 | x^k] \\ &= \mathbb{E}[\| \underbrace{h^{k-1}}_{\substack{\nabla f_{(i_k)}(x^*) \\ - \nabla f(x^*)}} + d(\underbrace{\nabla f_{(i_k)}(x^k)}_{\substack{\nabla f_{(i_k)}(x^*) \\ + \nabla f_{(i_k)}(x^*) - h_{(i_k)}^{k-1}}} - \underbrace{\nabla f_{(i_k)}(x^*)}_{\substack{\nabla f_{(i_k)}(x^*) \\ + \nabla f_{(i_k)}(x^*) - h_{(i_k)}^{k-1}}} - \underbrace{\nabla f(x^*)}_{\substack{\nabla f_{(i_k)}(x^*) \\ + \nabla f_{(i_k)}(x^*) - h_{(i_k)}^{k-1}}})\|_2^2 | x^k] \end{aligned}$$

КБЛ $\|a+b\|_2^2 \leq 2\|a\|_2^2 + 2\|b\|_2^2$, где a const

$$\begin{aligned} &\leq 2\mathbb{E}[\|d(\nabla f_{(i_k)}(x^k) - \nabla f_{(i_k)}(x^*))\|_2^2 | x^k] \\ &\quad + 2\mathbb{E}[\|d(h_{(i_k)}^{k-1} - \nabla f_{(i_k)}(x^*)) - (h^{k-1} - \nabla f(x^*))\|_2^2 | x^k] \end{aligned}$$

$$\mathbb{E} \gamma^2 \geq 10\gamma$$

$$\leq 2 \mathbb{E} [\|d(\nabla f_{(ik)}(x^k) - \nabla f_{(ik)}(x^*))\|_2^2 | x^k] \\ + 2 \mathbb{E} [\|d(h_{(ik)}^{k-1} - \nabla f_{(ik)}(x^*))\|_2^2 | x^k]$$

cm. rank b. property. lemma

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

L - property

$$= 4Ld (f(x^k) - f(x^*)) + 2d \|h^{k-1}\|_2^2 \quad (****)$$

Triangle inequality (****) & (****)

$$\mathbb{E} [\|x^{k+1} - x^*\|_2^2] = \mathbb{E} [\|x^k - x^*\|_2^2] \\ - 2\gamma \mathbb{E} [\frac{\mu}{2} \|x^k - x^*\|_2^2 + f(x^k) - f(x^*)] \\ + \gamma^2 (4Ld \mathbb{E} [f(x^k) - f(x^*)] + 2d \mathbb{E} [\|h^{k-1}\|_2^2]) \\ = (1 - \gamma\mu) \mathbb{E} [\|x^k - x^*\|_2^2] \\ - (2\gamma - 4\gamma^2 Ld) \mathbb{E} [f(x^k) - f(x^*)] \\ + 2d\gamma^2 \mathbb{E} [\|h^{k-1}\|_2^2] \quad (\diamond)$$

$\mathbb{E} [\|h^k\|_2^2]$ + tower property

$$\mathbb{E} [\|h^k\|_2^2 | x^k] = \mathbb{E} [\| \underline{h^{k-1}} + \underline{\nabla f_{(ik)}(x^k)} - \underline{h_{(ik)}^{k-1}} \|_2^2 | x^k]$$

$$\|a+b\|_2^2 = \|a\|_2^2 + \|b\|_2^2 + 2\langle a, b \rangle$$

$$= \mathbb{E} [\|h^{k-1} - h_{(ik)}^{k-1}\|_2^2 | x^k] + \mathbb{E} [\|\nabla f_{(ik)}(x^k)\|_2^2 | x^k] \\ + 2 \mathbb{E} [\langle \nabla f_{(ik)}(x^k); h^{k-1} - h_{(ik)}^{k-1} \rangle | x^k]$$

так в коры сумме

$$= \mathbb{E} [\|h^{k-1} - h_{(ik)}^{k-1}\|_2^2 | x^k] + \frac{1}{d} \|\nabla f(x^k)\|_2^2 \\ + 2 \mathbb{E} [\langle \nabla f_{(ik)}(x^k); h^{k-1} - h_{(ik)}^{k-1} \rangle | x^k] \equiv$$

$$2 \mathbb{E} [\langle \nabla f_{(ik)}(x^k); h^{k-1} - h_{(ik)}^{k-1} \rangle | x^k] \\ = 2 \mathbb{E} [\langle \nabla f_{(ik)}(x^k); h^{k-1} \rangle | x^k] + \langle \nabla f_{(ik)}(x^k); h_{(ik)}^{k-1} \rangle \\ - 2 \mathbb{E} [\langle \nabla f_{(ik)}(x^k); h_{(ik)}^{k-1} \rangle | x^k] = 0$$

$$\equiv \mathbb{E} [\|h^{k-1} - h_{(ik)}^{k-1}\|_2^2 | x^k] + \frac{1}{d} \|\nabla f(x^k)\|_2^2 \\ = \cancel{\mathbb{E} [\|h^{k-1}\|_2^2 | x^k]} + \mathbb{E} [\|h_{(ik)}^{k-1}\|_2^2 | x^k] - 2 \mathbb{E} [\langle h^{k-1}; h_{(ik)}^{k-1} \rangle | x^k] \\ + \frac{1}{d} \|\nabla f(x^k)\|_2^2$$

так в коры сумме

$$= \|h^{k-1}\|_2^2 + \frac{1}{d} \|h^{k-1}\|_2^2 - 2 \mathbb{E} [\langle h_{(ik)}^{k-1}; h_{(ik)}^{k-1} \rangle | x^k] \\ = \|h^{k-1}\|_2^2 + \frac{1}{d} \|h^{k-1}\|_2^2 - 2 \mathbb{E} [\|h_{(ik)}^{k-1}\|_2^2 | x^k] + \frac{1}{d} \|\nabla f(x^k)\|_2^2$$

так в коры сумме

$$= \left(1 - \frac{1}{d}\right) \|h^{k-1}\|_2^2 + \frac{1}{d} \|\nabla f(x^k)\|_2^2$$

L-выражение

$$\leq \left(1 - \frac{1}{d}\right) \|h^{k-1}\|_2^2 + \frac{2L}{d} (f(x^k) - f(x^*))$$

Answer: (\diamond)

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2] &\leq (1-\gamma\mu) \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - (2\gamma - 4\gamma^2 L d) \mathbb{E}[f(x^k) - f(x^*)] \\ &\quad + 2d\gamma^2 \mathbb{E}[\|h^{k+1}\|_2^2] \end{aligned}$$

$$\mathbb{E}[\|h^k\|_2^2] \leq \left(1 - \frac{1}{d}\right) \mathbb{E}[\|h^{k-1}\|_2^2] + \frac{2L}{d} \mathbb{E}[f(x^k) - f(x^*)]$$

Convergence c. unnummeren M :

$$\begin{aligned} \mathbb{E}[\|x^{k+1} - x^*\|_2^2 + M \|h^k\|_2^2] &\leq (1-\gamma\mu) \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - (2\gamma - 4\gamma^2 L d) \mathbb{E}[f(x^k) - f(x^*)] \\ &\quad + 2d\gamma^2 \mathbb{E}[\|h^{k+1}\|_2^2] \\ &\quad + \left(1 - \frac{1}{d}\right) \mathbb{E}[M \|h^{k-1}\|_2^2] + \frac{2LM}{d} \mathbb{E}[f(x^k) - f(x^*)] \\ &= (1-\gamma\mu) \mathbb{E}[\|x^k - x^*\|_2^2] \\ &\quad - \left(2\gamma - 4\gamma^2 L d - \frac{2LM}{d}\right) \mathbb{E}[f(x^k) - f(x^*)] \\ &\quad + \left(1 - \frac{1}{d} + \frac{2d\gamma^2}{M}\right) \mathbb{E}[M \|h^{k-1}\|_2^2] \end{aligned}$$

$$M: \quad 1 - \frac{1}{d} + \frac{2d\gamma^2}{M} = 1 - \frac{1}{2d} \Rightarrow \boxed{M = 4d\gamma^2}$$

$$\gamma: \quad 2\gamma - 4\gamma^2 L d - \frac{2LM}{d} \geq 0$$

$$2\gamma - 4\gamma^2 L d - 8L\gamma^2 d \geq 0$$

$$1 - 6\gamma L d \geq 0$$

$$\boxed{\gamma \leq \frac{1}{6Ld}}$$

$$\begin{aligned}
&\leq (1-\gamma\mu) \mathbb{E}[\|x^k - x^*\|_2^2] \\
&\quad + \left(1 - \frac{1}{2d}\right) \mathbb{E}[M \|h^{k-1}\|_2^2] \\
&\leq \max\left((1-\gamma\mu); \left(1 - \frac{1}{2d}\right)\right) \mathbb{E}[\|x^k - x^*\|_2^2 + M \|h^{k-1}\|_2^2]
\end{aligned}$$

Сформулируем SEGA:

$$\boxed{\mathbb{E}\|x^{k+1} - x^*\|_2^2 \leq \max\left\{(1-\gamma\mu); \left(1 - \frac{1}{2d}\right)\right\}^{k+1} \mathbb{E}[\|x^0 - x^*\|_2^2 + M \|h^{-1}\|_2^2]}$$

До момента ε ($\gamma = \frac{1}{6Ld}$)

$$O\left(\left(d + \frac{dL}{\mu}\right) \log \frac{1}{\varepsilon}\right) \text{ итераций.}$$

⊙ не нужно в начале, не важно, какой шаг

⊕ можно выбрать шаг не заранее