

Стохастический фрагментный спуск

x_* - опт. значение (*)

$$f(x) := \mathbb{E}_{\zeta} f(x, \zeta) \rightarrow \min_{x \in \mathbb{R}^n} \quad (*)$$

SGD: stochastic gradient descent

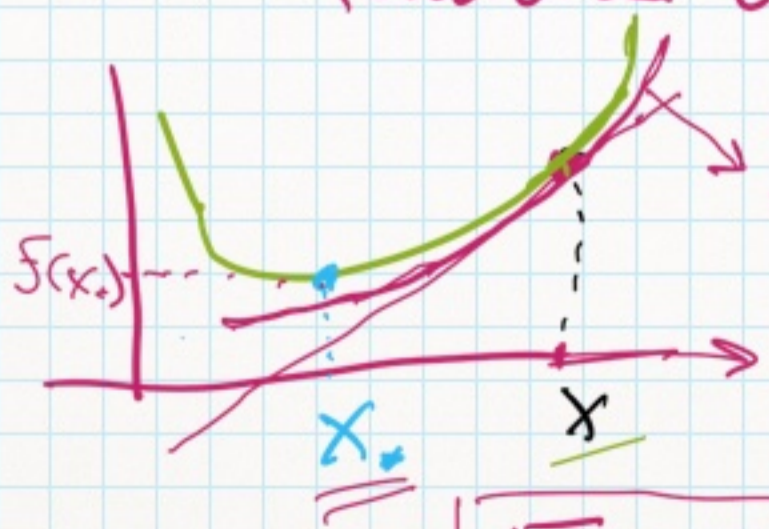
$$x^{k+1} = x^k - h \nabla_x f(x^k, \zeta^k), \quad \zeta^k \text{ i.i.d.}$$

Предположение 1. $\forall x \rightarrow \mathbb{E}_{\zeta} \nabla_x f(x, \zeta) = \nabla f(x)$

Предположение 2.

$$\mathbb{E}_{\zeta} [\|\nabla f(x, \zeta)\|_2^2] \leq 2L(f(x) - f(x_*)) + \sigma^2$$

Предположение 3. $f(x_*) \geq f(x) + \langle \nabla f(x), x_* - x \rangle + \frac{\mu}{2} \|x_* - x\|_2^2$
(сильная выпуклость)



$\frac{\mu}{2} \|x_* - x\|_2^2$

$(1 - \frac{\mu h}{2L})$

Теорема. $h \leq \frac{1}{2L}$

$$\mathbb{E} \|x^{k+1} - x_*\|_2^2 \leq (1 - \mu h)^k \|x^0 - x_*\|_2^2 + \frac{h\sigma^2}{\mu}$$

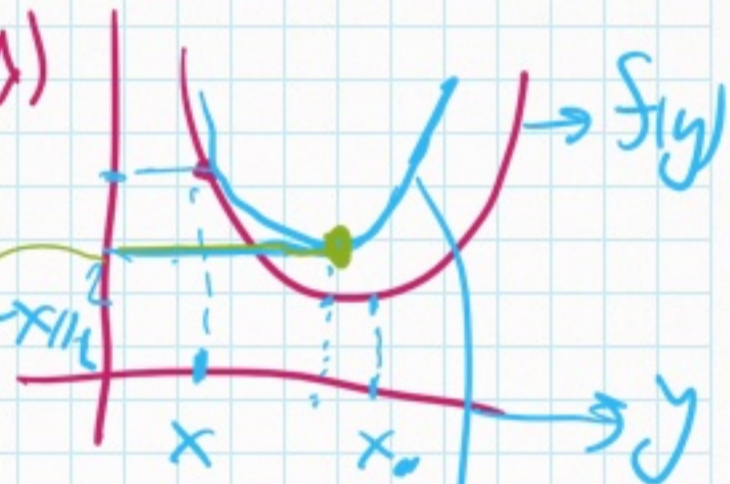
$$\mathbb{E}_{\xi} [\|\nabla f(x, \xi)\|_2^2] \leq 2L(f(x) - f(x_*)) + \sigma^2 \quad (**)$$

1 шаг (мин. шаг, нем. макс.)

$$\nabla f(x, \xi) \equiv \nabla f(x)$$

$$\| \nabla f(y) - \nabla f(x) \|_2 \leq L \|y - x\|_2 \quad (\Delta)$$

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



$$f(x) - \frac{1}{2L} \|\nabla f(x)\|_2^2$$

\(\forall\) $f(x_*)$

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

\(\min_y\)

$$\|\nabla f(x)\|_2^2 \leq 2L(f(x) - f(x_*))$$

$$\sigma^2 = 0$$

$$\mathbb{E} [\|\frac{1}{r} \sum_{\xi} \nabla f(x, \xi)\|_2^2] \leq \frac{\sigma^2}{r}$$

2 шаг $\nabla f(x, \xi) = \nabla f(x) + \xi$, $\mathbb{E} \|\xi\|_2^2 \leq \sigma^2$
 (***) - берем; L - конст. из (\Delta), σ^2

$$\mathbb{E}[\|\nabla f(x, \xi)\|_2^2] = \underbrace{\|\nabla f(x)\|_2^2}_{\substack{\text{recurr.} \\ \text{M. ber.}}} + \mathbb{E}\|\xi\|_2^2 \stackrel{!}{=} \sigma^2$$

$$\mathbb{E}\xi = 0$$

$$2L(f(x) - f(x_*))$$

3 случай

$$f(x) = \frac{1}{m} \sum_{i=1}^m f_i(x)$$

$\forall i=1, \dots, m$

$$\|\nabla f_i(y) - \nabla f_i(x)\|_2 \leq L\|y-x\|_2$$

$$\xi \in \mathbb{R}^{1, \dots, m}$$

$$\mathbb{E} f_\xi(x)$$

$$\mathbb{E}_\xi[\|\nabla f_\xi(x)\|_2^2] = \mathbb{E}_\xi \|\underbrace{\nabla f_\xi(x) - \nabla f_\xi(x_*)}_a + \underbrace{\nabla f_\xi(x_*)}_b\|_2^2$$

$$\leq 2\mathbb{E}_\xi \|\nabla f_\xi(x) - \nabla f_\xi(x_*)\|_2^2 + 2\mathbb{E} \|\nabla f_\xi(x_*)\|_2^2 \leq$$

$$(a+b)^2 \leq 2a^2 + 2b^2$$

$$\leq 4L(f(x) - f(x_*)) + 2\mathbb{E}_\xi[\|\nabla f_\xi(x_*)\|_2^2]$$

$$2L \nearrow$$

$\nabla f_i(x_*) = 0$
непроблемно

$$\frac{2}{m} \sum_{i=1}^m \|\nabla f_i(x_*)\|_2^2$$

$$f_i(x) = (y_i - g_i(x))^2$$

$$y_i = g_i(x) \quad i = \overline{1, m}$$

$$\sigma^2 = 0$$

Don-Go Teorema

$$h = h_k \quad x^{k+1} = x^k - h \nabla_x f(x^k, \xi^k)$$

Lemma

$$\mathbb{E} \|x^{k+1} - x_*\|_2^2 \leq (1 - \mu h) \mathbb{E} \|x^k - x_*\|_2^2 -$$

$$- h \mathbb{E} (f(x^k) - f(x_*)) + h^2 \sigma^2$$

$$h \leq \frac{1}{2L}$$

$$\textcircled{1} \mathbb{E} [\|x^{k+1} - x_*\|_2^2 \mid \underbrace{x^k, \dots, x^1}_{\mathcal{F}_k}, \underbrace{\xi^{k+1}, \dots, \xi^0}_{\mathcal{F}_k}] = \mathbb{E} [\|x^k - x_*\|_2^2 -$$

$$- 2h \langle \nabla_x f(x^k, \xi^k), x^k - x_* \rangle + h^2 \|\nabla_x f(x^k, \xi^k)\|_2^2 \mid \mathcal{F}_k] =$$

$$= \|x^k - x_*\|_2^2 - \underbrace{2h \langle \nabla f(x^k), x^k - x_* \rangle}_{\text{no M. condition}} + \text{no M. condition}$$

$$+ h^2 \mathbb{E} [\|\nabla_x f(x^k, \xi^k)\|_2^2 \mid \mathcal{F}_k] \leq \|x^k - x_*\|_2^2$$

$$\mathbb{E}_{\xi^k} [\|\nabla_x f(x^k, \xi^k)\|_2^2] \rightarrow \text{overubars}$$

opuka.

$$= 2h \left(\frac{\mu}{2} \|x^k - x_*\|_2^2 + (f(x^k) - f(x_*)) \right) +$$

$$+ h^2 (2L (f(x^k) - f(x_*)) + \sigma^2)$$

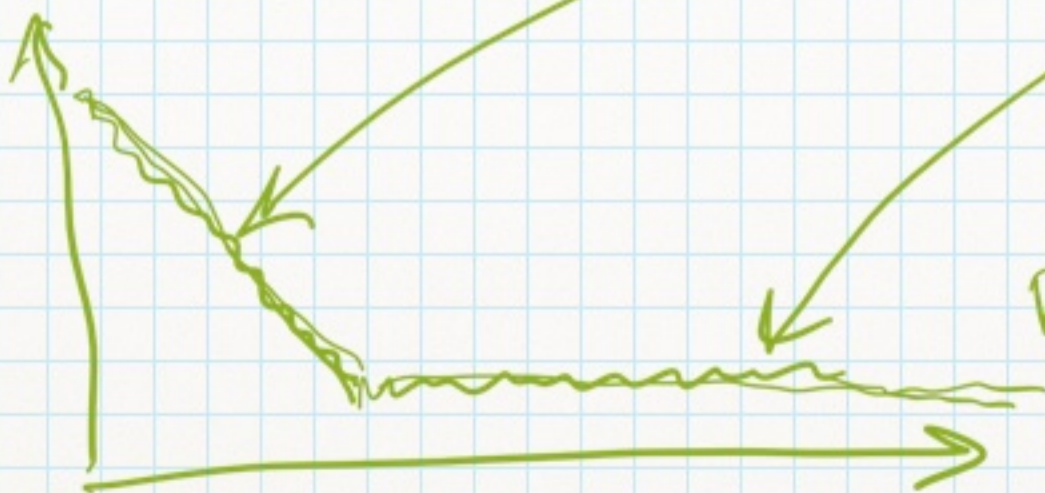
$$\mathbb{E}[\|x^{k+1} - x_*\|_2^2] \leq (1 - \mu h) \mathbb{E}[\|x^k - x_*\|_2^2] - \underbrace{2h(1 - Lh)}_{\mathbb{E} f(x^k) - f(x_*)} + h^2 \sigma^2$$

$$h \leq \frac{1}{2L} \Rightarrow (1 - Lh) \geq \frac{1}{2}$$

$$\mathbb{E}[\|x^{k+1} - x_*\|_2^2] \leq (1 - \mu h) \mathbb{E}[\|x^k - x_*\|_2^2] + h^2 \sigma^2$$

$$h \leq \frac{1}{2L} \quad \Downarrow \quad h^2 \sigma^2 \sum_{t=0}^{\infty} (1 - \mu h)^t = h^2 \sigma^2 \frac{1}{1 - (1 - \mu h)}$$

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



Mittlerwert

$$\mathbb{E} f(x, \mathcal{D}) \rightarrow \frac{1}{r} \sum_{j=1}^r f(x, \mathcal{D}_j)$$

$$\sigma^2 \rightarrow \sigma^2 / r$$

S. Stich'igs

$$E(f(\bar{x}^N)) - f(x_0) + \mu \underbrace{E \|x^N - x_0\|_2^2}_{\text{можно не знать}} =$$

$$\frac{1}{N} \sum_{k=0}^{N-1} x^k$$

$$= O\left(\min\left\{LR^2 \exp\left(-\frac{MN}{4L}\right) + \frac{\sigma^2}{MN},\right.\right.$$

$$\left.\frac{LR^2}{N} + \frac{\sigma R}{\sqrt{N}}\right\}$$

$$\boxed{\sigma^2 \rightarrow \frac{\sigma^2}{L}}$$

$$R = \|x^0 - x_0\|_2$$

связан ϵ_2^2

ϵ_2

$$= O\left(\min\left\{LR^2 \exp\left(-\text{const} \sqrt{\frac{M}{L} N}\right) + \frac{\sigma^2}{MN},\right.\right.$$

$Df(x) - L$ -линейн.
 $D^2f(x, z) - \sigma^2$ шум.

$$\frac{LR^2}{N^2} + \frac{\sigma R}{\sqrt{N}}$$

направит

ε - мом. по ошибке

$$\left(1 - \frac{M}{2L}\right)^N \cdot R^2 \approx \varepsilon$$

$$N \approx \frac{L}{M} \ln\left(\frac{R^2}{\varepsilon}\right) - \text{число измерений}$$

$$\frac{h\sigma^2}{M} \sim \frac{\sigma^2}{4M} \approx \varepsilon$$

$$\frac{\sigma^2}{4M} \approx \varepsilon \Rightarrow M \approx \frac{\sigma^2}{4\varepsilon}$$

$$x^{k+1} = x^k - h \underbrace{\nabla_x f(x^k, z^k)}_{\frac{1}{r} \sum_{j=1}^r \nabla_x f(x^k, z^{k,j})}$$

$$\frac{1}{r} \sum_{j=1}^r \nabla_x f(x^k, z^{k,j})$$

i.i.d.

$$\nabla f(x^k)$$

система
↓

Ускорение: $\left(1 - \sqrt{\frac{M}{L}}\right)^N R^2 = \varepsilon \Rightarrow N = \sqrt{\frac{L}{M}} \ln \frac{R}{\varepsilon}$

Weak growth condition

Vaswani
Bach, Schmidt, 13

$$\mathbb{E} [\|\nabla f(x, z)\|_2^2] \leq 2\rho L (\|f(x) - f(x_0)\|_1) \quad (\text{WGC})$$

Strong growth condition

$$\leq 2\rho L (f(x) - f(x_0)) \quad (\text{WGC})$$

$$\mathbb{E} [\|\nabla f(x, z)\|_2^2] \leq \rho \|\nabla f(x)\|_2^2 \quad (\text{SGC})$$

Укрепление условий

$$\mathbb{E} f(x^k) - f(x_0) \leq \left(1 - \sqrt{\frac{M}{\rho^2 L}}\right)^k (f(x^0) - f(x_0) + \frac{M}{2} \|x^0 - x_0\|_2^2)$$

dim x

$$n \langle \nabla f(x), e \rangle e = \rho f(x, e)$$

$e \in S^n(1)$ — сфера, радиус 1

$$\mathbb{E} \nabla f(x, e) = \nabla f(x)$$

$$\rho = n \quad \text{SGC}$$

$$\left(1 - \frac{1}{n} \sqrt{\frac{M}{L}}\right)^N \Delta f = \varepsilon$$

$$N \sim n \sqrt{\frac{1}{\rho} \ln \frac{1}{\varepsilon}}$$

Fang, Fan, Friedlander. ICLR 2021

$$f(x) := \frac{1}{m} \sum_{i=1}^m f_i(x)$$

$$\nabla f_{\zeta}(x) = \underbrace{\ell'(h(x))}_{\wedge_m} \cdot \underbrace{\nabla h(x)}_M$$

$$\| \nabla h(x) \|_2 \leq M$$

↙ $\ell(h(x))$, $\ell: \mathbb{R} \rightarrow \mathbb{R}_+$

$$| \ell'(\alpha) - \ell'(\beta) | \leq L |\alpha - \beta| \quad (L = \infty)$$

$$\ell(0) = 0, \quad \ell(\alpha) \geq 0$$

↘ $\ell'(0) = 0$

$$|h(y) - h(x)| \leq M \|y - x\|_2$$

$$\| \nabla_x f(x, \zeta) \|^2 = \| \nabla f_{\zeta}(x) \|^2 \leq M^2 (\ell'(h(x)))^2 =$$

$$= M^2 (\ell'(h(x)) - \ell'(0))^2 \leq 2M^2 (\ell(h(x)) - \ell(0)) =$$

$$= 2M^2 f_{\zeta}(x)$$

$$(L = \infty) \left[\frac{1}{2L} (\ell'(h))^2 \leq \ell(h) - \ell(0) \right]$$

↙ $L=1$

$$\| \nabla_x f(x, \zeta) \|^2 \leq 2M^2 (f_{\zeta}(x) - f(x_*)) + 2M^2 f(x_*)$$

$$\mathbb{E} \| \nabla_x f(x, \zeta) \|^2 \leq 2M^2 (f(x) - f(x_*)) + \underbrace{2M^2 f(x_*)}_{\sigma^2}$$