

Менюшев  
Гасников  
Александр  
@gasnikov

gasnikov@yandex.ru

Градиентный  
спуск

Г. Коши

$$\min_x f(x)$$



$$\frac{dx}{dt} = -\nabla f(x) \quad \left\{ \begin{array}{l} \text{система Коши} \\ x(t+1) \rightarrow x_{loc} - \text{лок.} \\ \text{мин.} \\ \text{(Турбина)} \end{array} \right.$$

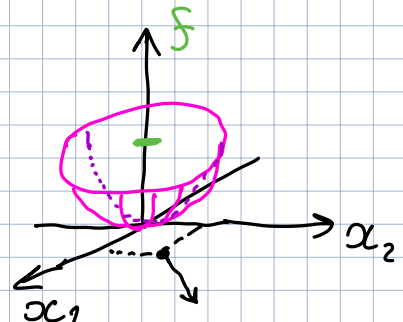
Пример.

$$f(x_1, x_2) = x_1^2 + x_2^2$$

$$f(x_1, x_2) = \text{const}$$

$$x_1^2 + x_2^2 = \text{const}$$

Если  $x_1 = \text{const}$   
 $f(\text{const}, x_2) = \text{const}^2 + x_2^2$



$$\nabla f(x_1, x_2) = (2x_1, 2x_2)^T \left\{ \left( \frac{\partial f}{\partial x_1}, \frac{\partial f}{\partial x_2} \right)^T \right.$$

$$\frac{dx}{dt} = l \Rightarrow \begin{cases} \frac{dx_1}{dt} = l_1 \\ \frac{dx_2}{dt} = l_2 \end{cases}$$

$$\begin{aligned} \frac{df}{dt}(x_1, x_2) &= \\ &= \frac{df}{dx_1} \frac{dx_1}{dt} + \frac{df}{dx_2} \frac{dx_2}{dt} = \end{aligned}$$

$$= \langle \nabla f(x), l \rangle = \|\nabla f(x)\|_2^2$$

↓  
 $\nabla f(x)$

$$\frac{df}{dt}(x_1, x_2) \stackrel{l = -\nabla f(x)}{=} -\langle \nabla f(x), \nabla f(x) \rangle = -\|\nabla f(x)\|_2^2$$

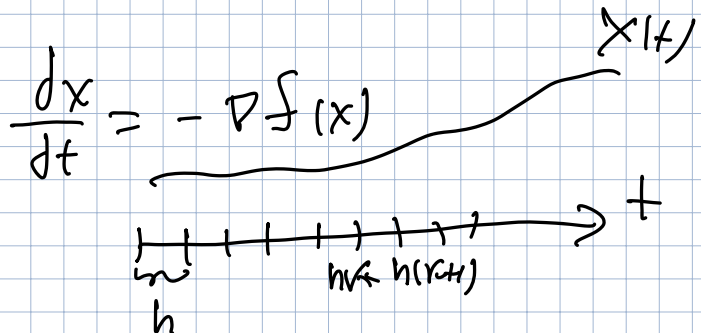
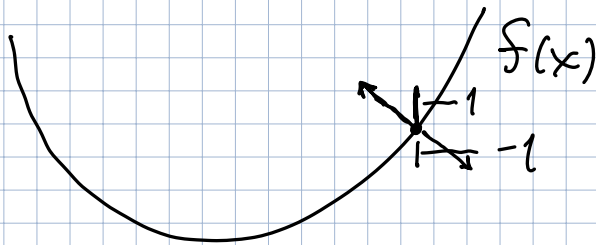


схема Эйлера

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

размерности  
случае

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

$[f]_{\frac{p \times q}{m}} =$   
 $= m.$   
 $\downarrow$   
 $\Sigma(?) = \frac{m^2}{p \times q}$

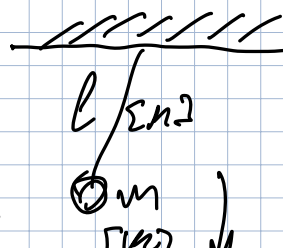
$M \times 1$   $f(x)$   
 $x$   
 $\nabla f(x)$   
 $\|x^0 - x^* \|_2 \leq R$

$\forall x, y$   $|x|, m=1$   
 $|f(y) - f(x)| \leq M \|y - x\|_2$   
 $\|\nabla f(x)\|_2 \leq M$   
 $\|\nabla f(y) - \nabla f(x)\|_2 \leq L \|y - x\|_2$   
 $[f]_{\frac{p \times q}{m}}$   
 $|x|, L=C$

|                                 |   |   |                                       |
|---------------------------------|---|---|---------------------------------------|
| $x \in \mathbb{R}^m$            | } | <del><math>\mathbb{R} \in \mathbb{R}^1</math></del> |                                       |
| $f \in \mathbb{R}^{p \times q}$ |   | $M \in \mathbb{R}^{p \times q / m}$                 | $h \in \mathbb{R}^{m^2 / p \times q}$ |
|                                 |   | $L \in \mathbb{R}^{p \times q / m^2}$               |                                       |
|                                 |   | $\varepsilon \in \mathbb{R}^{p \times q}$           |                                       |

$f(x^N) - f(x^*) \leq \varepsilon$  — точность по  $\varepsilon$ -уровню

$\Pi$ -теорема теории размерности



задача о мультипликативности (Ганделс)

$T \in \mathbb{C}^{m \times n}$   
 $T / \sqrt{e/g}$   
 $\frac{[C \times C]}{\sqrt{m / (m \times C \times C)}} = \frac{[C \times C]}{[C \times C]} = g \in \mathbb{R}^{m / C \times C}$

$\Pi$ -Теорема

$$F(T/\sqrt{g/l}) \equiv 1$$

$$T/\sqrt{g/l} = F^{-1}(1) = \text{const}$$

$$T = \text{const} \sqrt{\frac{l}{g}}$$

$$\Sigma \Sigma \rho y \delta z$$

$$M \Sigma \rho y \delta / r^2$$

$$h \Sigma v r^2 / \rho y \delta z$$

$$\frac{h}{E/M^2} = \frac{v r^2 / \rho y \delta z}{\rho y \delta z / (\rho y \delta / r^2)^2}$$
$$= \frac{v r^2 / \rho y \delta z}{v r^2 / \rho y \delta z} = 1$$

$\text{const} \equiv 1$

}

$$h = \text{const} + \frac{\Sigma}{M^2}$$

}

субординированности  
матрицы

$$L \Sigma \rho y \delta / v r^2$$

$\text{const} \equiv 1$

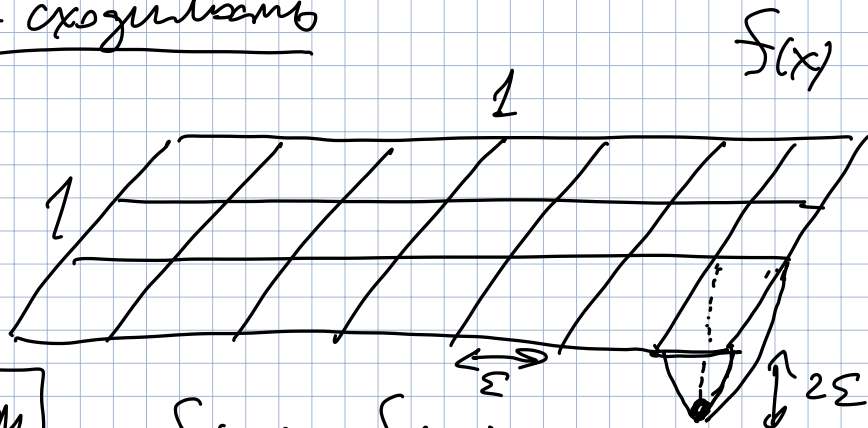
}

$$h = \text{const} \cdot \frac{1}{L}$$

$$x^{k+1} = x^k - \frac{\Sigma}{L^2} \nabla f(x^k)$$

$$x^{k+1} = x^k - \frac{\Sigma}{\|\nabla f(x^k)\|_2^2} \nabla f(x^k)$$

Плохая сходимость



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

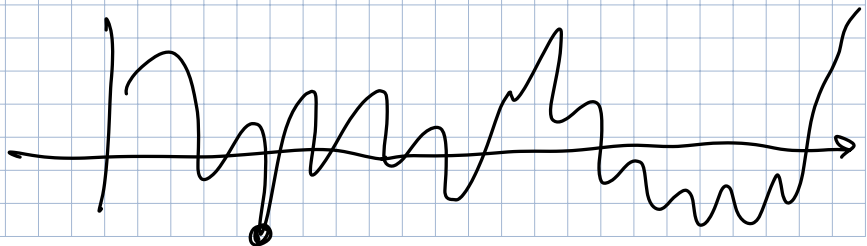
$$f(x_1) - f(x_0) \leq \epsilon$$

итераций  $\sim \frac{1}{\epsilon^2}$

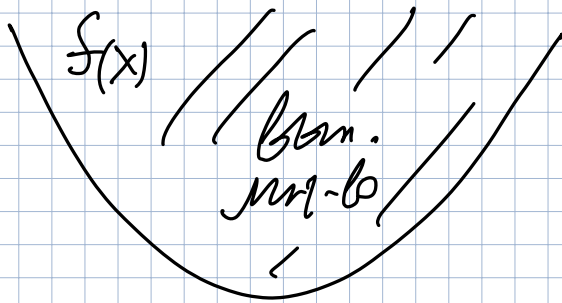
$$\min_{x \in \mathbb{R}^2} f(x)$$

итераций  $\sim \frac{1}{\epsilon}$

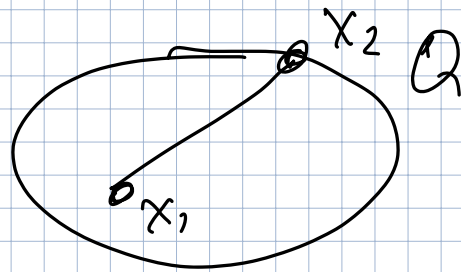
$$\min_{x \in \mathbb{R}^d} f(x)$$



# Вспомогательная лемма.



$\min_{x \in Q} f(x)$   
 $\downarrow$  блн.



$\forall x_1, x_2 \in Q \Rightarrow$   
 $\Rightarrow [x_1, x_2] \in Q$

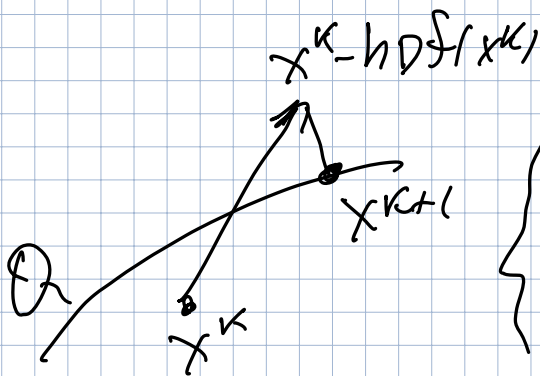
$\forall x \in Q$

$$\|\nabla f(x)\| \leq M$$

$$\|x^0 - x\|_2 \leq R$$

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

$\nearrow$   $\pi_Q$   
 проекция на множество Q  
 $\downarrow$   $\frac{\varepsilon}{M^2}$



$$= \operatorname{argmin}_{x \in Q} \left\{ h \nabla f(x^k), \right.$$

$$\left. x - x^k > + \frac{1}{2} \|x - x^k\|_2^2 \right\}$$



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

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

$$\Pi_Q (x^k - h \nabla f(x^k)) - x_*$$

$$\Pi_Q (x^k - h \nabla f(x^k) - x_*)$$

$$\|\Pi_Q\|_2 \leq 1$$

$$\|\Pi_Q(y)\|_2 \leq \|y\|_2 \quad (*)$$

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

$$+ h^2 \underbrace{\|\nabla f(x^k)\|_2^2}_{M^2}$$

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

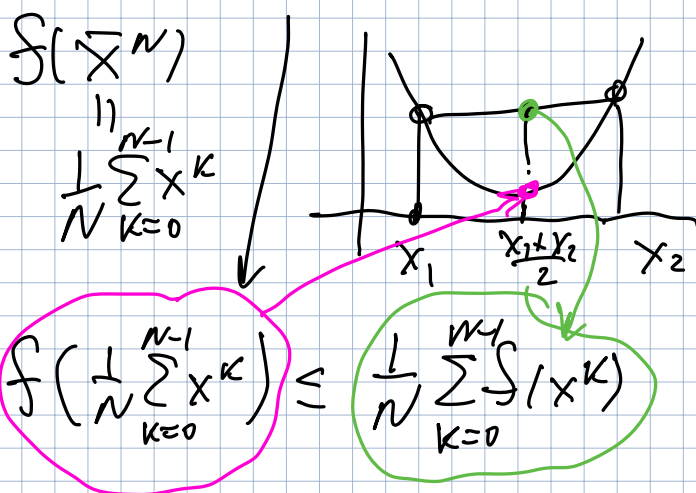


$$\underbrace{\langle \nabla f(x^k), x^k - x_0 \rangle}_{\substack{\text{V. } \text{бun.} \\ f(x^k) - f(x_0)}} \leq \frac{1}{2h} \|x^k - x_0\|^2 - \frac{1}{2h} \|x^{k+1} - x_0\|^2 + \frac{h}{2} M^2$$

$$\frac{1}{N} \sum_{k=0}^{N-1} \left\{ f(x^k) - f(x_0) \leq \frac{1}{2h} \|x^k - x_0\|^2 - \frac{1}{2h} \|x^{k+1} - x_0\|^2 + \frac{h}{2} M^2 \right.$$

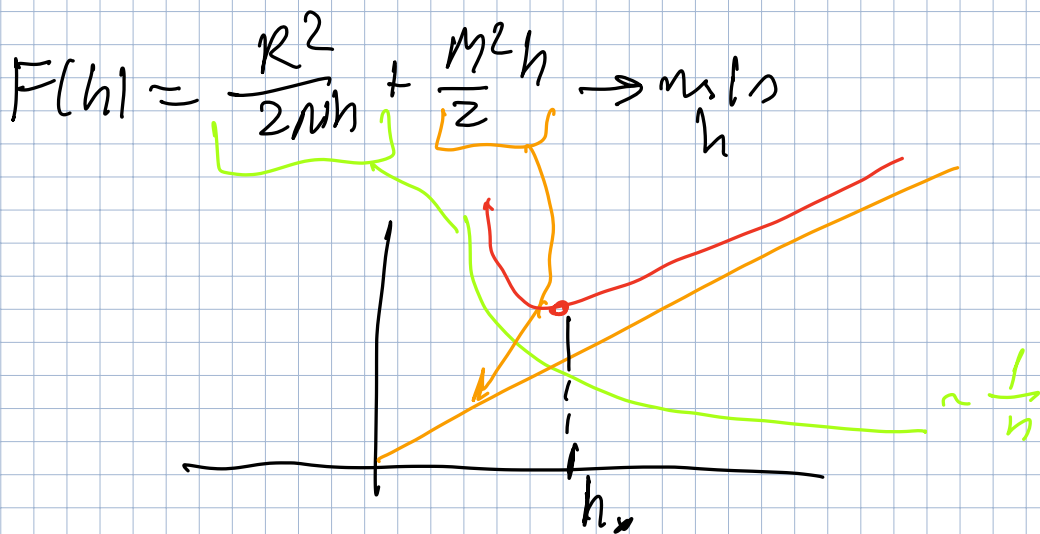
$$\underbrace{\frac{1}{N} \sum_{k=0}^{N-1} f(x^k) - f(x_0)}_{\substack{\text{V. } \text{н-бо} \\ \text{Уменьша}}}} \leq \frac{1}{2hN} R^2 - \frac{1}{2h} \|x^N - x_0\|^2 + \frac{h}{2} M^2$$

$\nearrow \|x^0 - x_0\|^2$



$$f(\bar{x}^n) - f(x_0) \leq \underbrace{\frac{R^2}{2Nh} + \frac{M^2 h}{2}}_{\text{beim } z \text{ und } h \rightarrow 0}$$

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



$$F'(h) = 0$$

$$-\frac{R^2}{2Nh^2} + \frac{M^2}{2} = 0$$

$$h_* = \frac{R}{M\sqrt{N}}$$

$$F(h_*) = \frac{MR}{\sqrt{N}}$$

$\min_{x \in Q} f(x)$   
 $\|\nabla f(x)\|_2 \leq M$   
 $\|x^0 - x_*\|_2 \leq R$

$$f(\bar{x}^N) - f(x_*) \leq \frac{MR}{\sqrt{N}} = \varepsilon$$

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

$$N = \frac{M^2 R^2}{\varepsilon^2}$$

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

$$h = \frac{R}{M\sqrt{N}} = \frac{\varepsilon}{M^2}$$

$$f(\bar{x}^N) - f(x_*) \leq \frac{MR}{\sqrt{N+1}}$$

Лекция 8

Ускоренный  
градиентный  
спуск

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

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

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

$$f(x) = x^2$$

$$L_f = ?$$