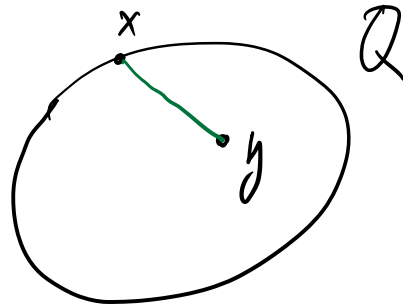
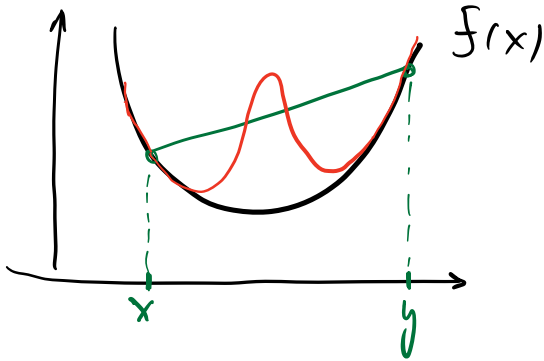
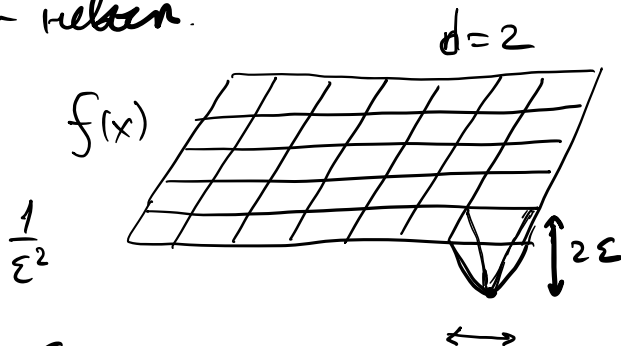


Лекция 1. Методы разрешимости Тунга

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



$f(x)$ - решение.

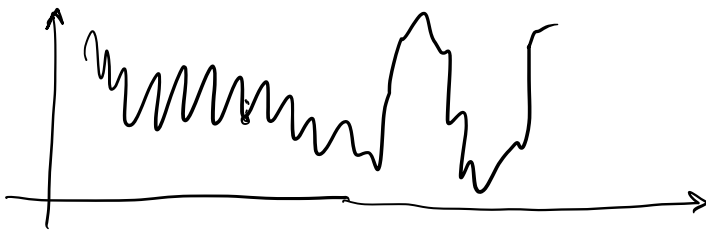


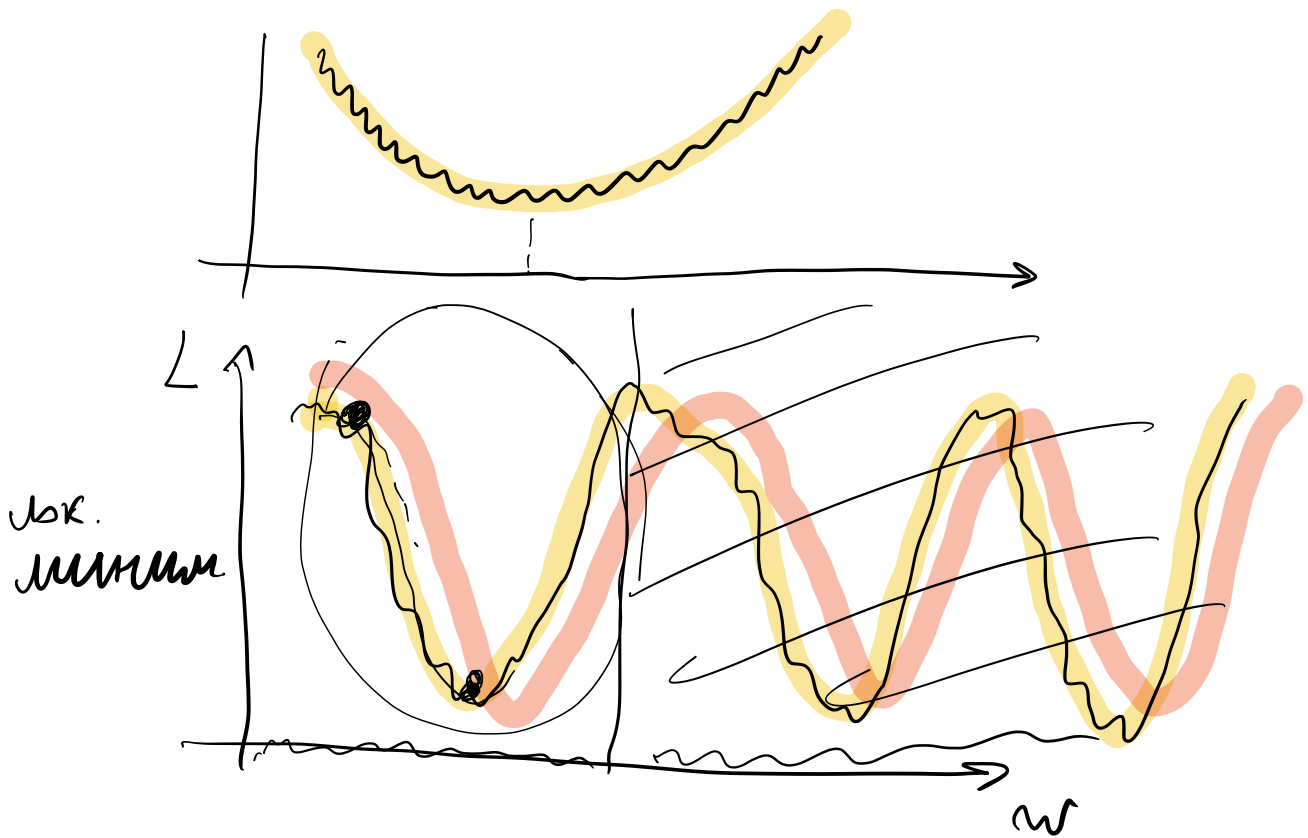
$\epsilon > 0$ - точность
решения
на ϵ -уровне
 $f(x^N) - f(x_*) \leq \epsilon$
 x_* - опт. воз.

$$1) |f(y) - f(x)| \leq M \|y - x\|_2 \quad \dots \sim \epsilon$$

$$2) \|\nabla f(y) - \nabla f(x)\|_2 \leq L \|y - x\|_2 \quad \dots \sim \sqrt{\epsilon}$$

$$\left(\frac{1}{\epsilon}\right)^d - \text{число итераций}$$





Выпуклая оптимизация

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

Q - область определения

$$Q = \{x \in \mathbb{R}^d \mid x \geq 0\}$$

$$x^{k+1} = x^k - h \nabla f(x^k) \quad Q = \mathbb{R}^d$$

$$x^{k+1} = \underbrace{\Pi_Q}_{\text{проект.}} \left(x^k - \underbrace{h \nabla f(x^k)}_{\text{граде}} \right) \quad A = d \begin{bmatrix} d \\ \end{bmatrix}$$

$$f(x) = \frac{1}{2} \langle x, Ax \rangle - \langle b, x \rangle$$

$$\nabla^2 f(x) = O(d^2)$$

$$\text{Εάν } Q = \mathbb{R}_+^d$$

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

norm. step $O(d)$

$f(x)$ -conv, Q -conv

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

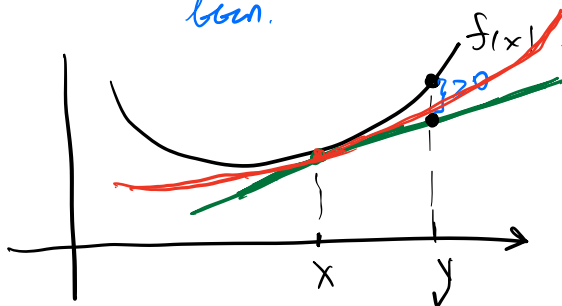
1) $|f(y) - f(x)| \leq M \|y - x\|_2$

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

$f(x)$ - μ -convex βαν. φ. υ. σ.

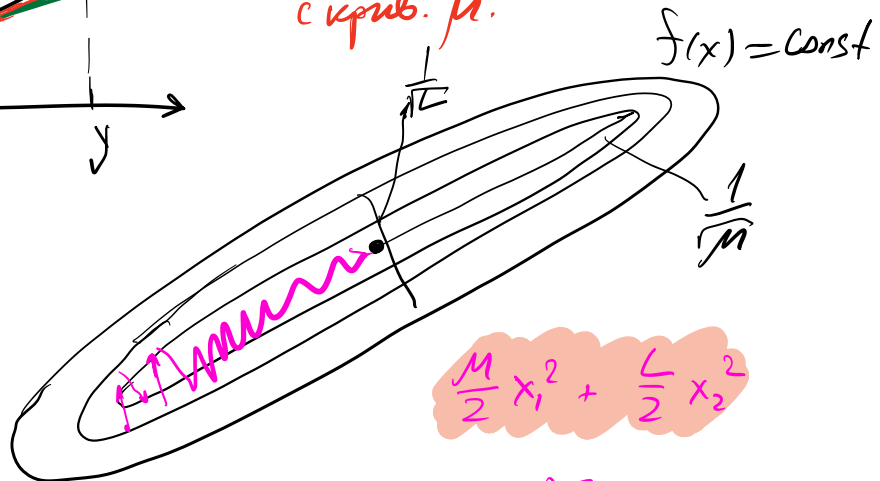
$$f(y) \geq f(x) + \langle \nabla f(x), y - x \rangle + \frac{\mu}{2} \|y - x\|_2^2$$

βαν.



καρβ. σ κρυσ. μ.

Το R είναι η απόσταση
 $R = \|x^0 - x_*\|_2$
 η μέση απόσταση
 και βολή x^0



$$\frac{M}{2} x_1^2 + \frac{L}{2} x_2^2$$

$$L = \max_x \lambda_{\max}(\nabla^2 f(x))$$

$$m = \min_x \lambda_{\min}(\nabla^2 f(x))$$

Курсов. по А.С. Хемпбери. '73

1) $N \leq d$ $\frac{\varepsilon}{M^2}$ $\sqrt{|x|}$ $M=1$

Субградиент. $x^{k+1} = x^k - h \nabla f(x^k)$ } $Q = \mathbb{R}^d$
 Метод.

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

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

$N \approx \frac{M^2 R^2}{\varepsilon^2}$ } где α $\sim \left(\frac{1}{\varepsilon}\right)^\alpha$

$x^{k+1} = x^0 + \text{Lin} \{ \nabla f(x^0), \nabla f(x^1), \dots, \nabla f(x^k) \}$

$f(p, y, \delta)$	$M [p, y, \delta / \kappa]$
$x [\kappa, \kappa]$	$R [\kappa]$
	$\varepsilon [p, y, \delta]$

$N = \left(\frac{MR}{\varepsilon}\right)^\alpha, \alpha = 2$

$x^{k+1} = x^k - h \nabla f(x^k)$
 $\kappa_2 = \kappa_2 - \underbrace{[?] \cdot \frac{p, y, \delta}{\kappa_2}}_{\kappa_2}$

$h \left[\frac{\kappa_2^2}{p, y, \delta} \right]$

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

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

Общеп.
ураж.
методы.

Несмеще
(усл., усложн.)

$$N \approx \sqrt{\frac{LR^2}{\varepsilon}}$$

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

$f(x)$ - μ -сильно вып.

$$1) \quad N \approx \frac{M^2}{\mu \varepsilon}$$

$$2) \quad N \approx \sqrt{\frac{L}{\mu}} \ln \left(\frac{MR^2}{\varepsilon} \right)$$

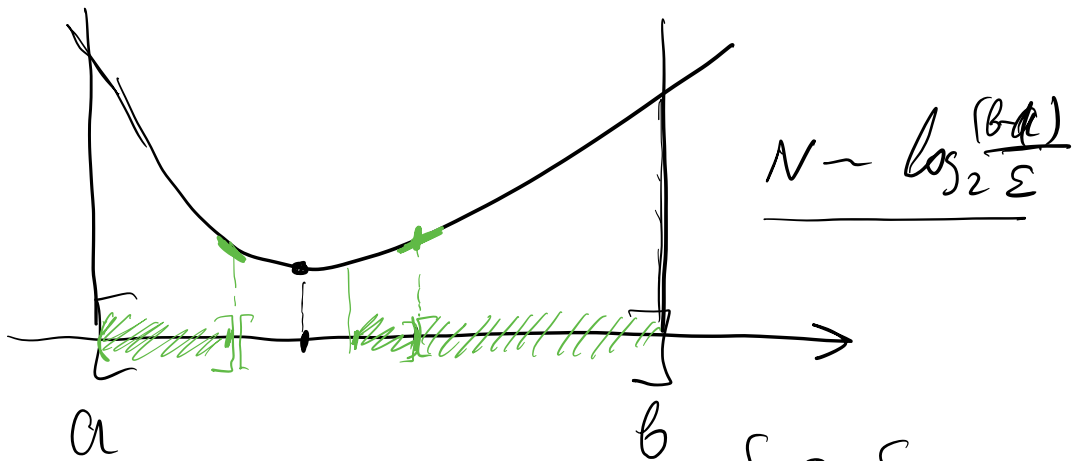
$\nabla f \rightarrow$	N $N \leq \delta$	$ f(y) - f(x) \leq M \ y - x\ _2$	$\ \nabla f(y) - \nabla f(x)\ _2 \leq L \ y - x\ _2$
$f(x)$ - вып.		$\frac{M^2 R^2}{\varepsilon^2}$	$\sqrt{\frac{LR^2}{\varepsilon}}$
$f(x)$ - μ -сильно вып.		$\frac{M^2}{\mu \varepsilon}$	$\sqrt{\frac{L}{\mu}} \ln \frac{MR^2}{\varepsilon}$

$$N \leq \delta$$

А там $N \geq \delta$

$$\min_{x \in [a, b]} f(x)$$

$$\underline{\delta = 1}$$



$$N \approx d \cdot \ln \left(\frac{\Delta f}{\epsilon} \right)$$

$f(x^0), f(x_*)$

$$\left[\nabla f(x), O(d^3) \right]$$

Доказательство (□)

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

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

$$V(x) = \frac{1}{2} \|x - x_*\|_2^2 \quad \text{— г-условная функция}$$

Случайные шаги: $\frac{dx}{dt} = -\nabla f(x)$ }

Схема Эулера

Если $f(x)$ — вып. $x_* \in Q$

$$\frac{dV}{dt} = \langle \nabla V(x), \frac{dx}{dt} \rangle =$$

$$= \langle \nabla f(x), x_* - x \rangle \leq \leftarrow \text{no term.}$$

$$\leq f(x_*) - f(x) \leq 0$$

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

$$\leq \|x^k - h \nabla f(x^k) - x_0\|_2^2 =$$

$$= \|x^k - x_0\|_2^2 - 2h \langle \nabla f(x^k), x^k - x_0 \rangle + \underbrace{h^2 \|\nabla f(x^k)\|_2^2}_{\leq M^2}$$

$$\|x^{k+1} - x_0\|_2^2 \leq \|x^k - x_0\|_2^2 - 2h \langle \nabla f(x^k), x^k - x_0 \rangle + h^2 M^2$$

$$2h (f(x^k) - f(x_0)) \leq 2h \langle \nabla f(x^k), x^k - x_0 \rangle \leq$$

← *by m.*

$$f(x_0) \geq f(x^k) + \langle \nabla f(x^k), x_0 - x^k \rangle$$

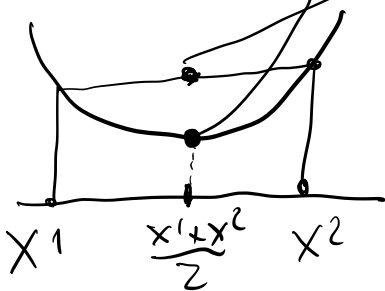
$$\leq \|x^k - x_0\|_2^2 - \|x^{k+1} - x_0\|_2^2 + h^2 M^2 \quad \left. \begin{array}{l} \sum_{k=0}^{N-1} \\ R^2 \end{array} \right\}$$

$$0 \leq 2h \sum_{k=0}^{N-1} (f(x^k) - f(x_0)) \leq (\|x^0 - x_0\|_2^2) - \|x^N - x_0\|_2^2 + N h^2 M^2 \quad \left. \begin{array}{l} \\ \frac{1}{N} \end{array} \right\}$$

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

VI H-b. Jensen

$$f\left(\frac{1}{N} \sum_{k=0}^{N-1} x^k\right)$$



$$2h (f(\bar{x}^N) - f(x_0)) \leq \frac{R^2}{N} + h^2 M^2$$

$$\|x^N - x_0\|_2^2 \leq R^2 + \underbrace{N h^2 M^2}_{R^2} \quad \text{h Good}$$

$$\|x^N - x_0\|_2^2 \leq 2R^2$$

$$f(\bar{x}^N) - f(x_0) \leq \frac{R^2}{2Nh} + \frac{hM^2}{2} \quad \forall h \geq 0$$

$$\min_h \left(\frac{R^2}{2Nh} + \frac{hM^2}{2} \right) \rightarrow \psi(h)$$

$$\psi'(h) = 0$$

$$-\frac{R^2}{2Nh^2} + \frac{M^2}{2} \Rightarrow h = \frac{R}{M\sqrt{N}}$$

$$\frac{R^2 M \sqrt{N}}{2NR} + \frac{M^2 R}{2M\sqrt{N}}$$

$$\frac{MR}{\sqrt{N}}$$

$$\frac{MR}{\sqrt{N+1}}$$

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

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

$$h = \frac{R}{M\sqrt{N}}$$

$$\Rightarrow N = \frac{M^2 R^2}{\epsilon^2}$$

$$M_k \approx \|\nabla f(x^k)\|_2$$

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

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

$$|f(y) - f(x)| \leq M \|y - x\|_2 ;$$

$$M = \max_{x \in Q} \|\nabla f(x)\|_2$$