

09.10.17 моно VI

Квазиинтерполирование методом QN

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

$$f(x_k + d) \approx m_k(d) = f_k + \nabla f_k^T d + \frac{1}{2} d^T B_k d \rightarrow \min_d$$

$$\nabla_d m_k(d) = \nabla f_k + B_k d = 0 \Rightarrow d = -B_k^{-1} \nabla f_k$$

$$x_{k+1} = x_k - \alpha_k B_k^{-1} \nabla f_k$$

условие секущей:
$$\begin{cases} \nabla m_{k+1}(0) = \nabla f_{k+2} \\ \nabla m_{k+2}(-\alpha_k d_k) = \nabla f_k \end{cases}$$

$$\nabla m_{k+1}(d) = \nabla f_{k+2} + B_{k+1} d$$

$$\nabla m_{k+1}(0) = \nabla f_{k+2} \quad \forall B_{k+1}$$

$$\nabla m_{k+2}(-\alpha_k d_k) = \nabla f_{k+2} + B_{k+2} (-\alpha_k d_k) = \nabla f_k$$

$$\Rightarrow B_{k+2} \underbrace{(x_{k+2} - x_k)}_{s_k} = \underbrace{\nabla f_{k+2} - \nabla f_k}_{y_k} \Rightarrow B_{k+2} s_k = y_k \mid s_k^T x$$

$$s_k^T B_{k+2} s_k = s_k^T y_k \Rightarrow s_k^T y_k > 0$$

умб достаточные условия

① Если f - строго выпукла, то $s_k^T y_k > 0$

② $\alpha_k^T \nabla f_k < 0$ и α_k выбирается из условия Золотого

$$\forall \alpha \in (0, 1) \Rightarrow s_k^T y_k > 0$$

Угера QN: $B_{k+1} = B_k + \text{low-rank-update}(B_k, s_k, y_k)$

Symmetric Rank 1 (SR1)

$$B_{k+1} = B_k + \sigma_k v_k v_k^T$$

$$B_{k+1} s_k = B_k s_k + \sigma_k v_k (v_k^T s_k) = y_k$$

$$\sigma_k (v_k^T s_k) v_k = y_k - B_k s_k \Rightarrow v_k = \sigma_k (y_k - B_k s_k)$$

$$\sigma_k \sigma_k^2 (y_k - B_k s_k)^T s_k (y_k - B_k s_k) = y_k - B_k s_k$$

$$\sigma_k \sigma_k^2 (y_k - B_k s_k)^T s_k = 1$$

$$\begin{aligned} B_{k+1}^{SR1} &= B_k + \sigma_k \sigma_k^2 (y_k - B_k s_k) (y_k - B_k s_k)^T = \\ &= B_k + \frac{(y_k - B_k s_k) (y_k - B_k s_k)^T}{(y_k - B_k s_k)^T s_k} \end{aligned}$$

$$d_k = -B_k^{-1} \nabla f_k$$

$$C_k = B_k^{-1}, \text{ можно использовать Бунгеппу:}$$

$$C_{k+1}^{SR1} = C_k + \frac{(C_k y_k - s_k) (C_k y_k - s_k)^T}{(C_k y_k - s_k)^T y_k}$$

Схема SR1

x_0, ε

$$d_0 = -\nabla f(x_0)$$

для $k=0, 1, 2, \dots$

$$d_k = \arg \min_d f(x_k + d d_k) \leftarrow \text{непочная}$$

$$x_{k+1} = x_k + d_k d_k$$

$$\nabla f_{k+1} = \nabla f(x_{k+1}), \text{ если } \|\nabla f_{k+1}\|^2 < \varepsilon, \text{ то стоп}$$

$$s_k = x_{k+1} - x_k, y_k = \nabla f_{k+1} - \nabla f_k$$

$$C_{k+1}^{SR1} = C_k + (\dots)$$

$$d_{k+1} = -C_{k+1}^{SR1} \nabla f_{k+1}$$

$$f(x) = \frac{1}{2} x^T A x - x^T b \rightarrow \min_x, \quad A > 0$$

$$\underline{\text{умб}} \quad \forall x_0, c_0 > 0, (C_k y_k - s_k)^T y_k \neq 0$$

$$\Rightarrow \text{в SR1 для } f(x) \text{ верно } C_k y_i = s_i \quad \forall i < k$$

$$\nabla f(x) = Ax - b$$

$$\nabla f_{k+2} - \nabla f_k = A(x_{k+2} - x_k) \Rightarrow y_k = A s_k$$

$$] [s_0 | \dots | s_{n-1}] \text{ л.н.г. : } y_k = A s_k \quad \forall k=0, \dots, n-1$$

$$A = [y_0 | \dots | y_{n-1}] [s_0 | \dots | s_{n-1}]^{-1}$$

$$] B_k > 0$$

$$B_{k+2}^{SR2} = B_k + \frac{(y_k - B_k s_k)(y_k - B_k s_k)^T}{(y_k - B_k s_k)^T s_k}$$

$$x^T B_{k+2}^{SR2} x = \underbrace{x^T B_k x}_0 + \frac{x^T (y_k - B_k s_k)(y_k - B_k s_k)^T x}{\underbrace{y_k^T s_k - s_k^T B_k s_k}_0} \neq 0$$

$$\tilde{B}_{k+1}^{SR2} = \begin{cases} B_k, & \text{если } (y_k - B_k s_k)^T s_k \leq \rho \|y_k - B_k s_k\| \|s_k\| \\ B_k^{SR2}, & \text{иначе} \end{cases} \quad \rho = 10^{-8}$$

[пропуска обновления]

Выбор C_0

$$C_0 = \delta_0 I, \quad \hat{x}_1 = x_0 - C_0 \nabla f(x_0)$$

$$s_0 = \hat{x}_1 - x_0, \quad y_0 = \nabla f(\hat{x}_1) - \nabla f(x_0)$$

$$\| \delta_0 y_0 - s_0 \|^2 \rightarrow \min_{\delta_0}, \quad \delta_0^2 y_0^T y_0 - 2\delta_0 y_0^T s_0 + s_0^T s_0 \rightarrow \min_{\delta_0}$$

$$\delta_0 = \frac{y_0^T s_0}{y_0^T y_0}$$

BFGS

$$\text{Хотим: } B_{k+1} s_k = y_k, \quad B_{k+1} > 0$$

$$B_{k+1} = J_{k+1} J_{k+1}^T \quad \text{параметризация}$$

$$B_{k+1} s_k = y_k \Leftrightarrow J_{k+1} \underbrace{J_{k+1}^T s_k}_{u_k} = y_k \Leftrightarrow \begin{cases} J_{k+1} s_k = u_k \\ J_{k+1} u_k = y_k \end{cases}$$

$$B_k = J_k J_k^T : \begin{cases} \frac{1}{2} \|J_{k+1} - J_k\|^2 \rightarrow \min_{J_k} \\ J_{k+1} s_k = u_k \\ J_{k+1} u_k = y_k \end{cases}$$

$$\mathcal{L}(y_{k+1}, \lambda) = \frac{1}{2} \| (y_{k+1} - y_k)^T (y_{k+1} - y_k) \| + \lambda^T (y_{k+1}^T s_k - u_k)$$

$$\nabla_{y_{k+1}} \mathcal{L} = y_{k+1} - y_k + s_k \lambda^T = 0, \quad y_{k+1} = y_k - s_k \lambda^T$$

$$y_{k+1}^T s_k = y_k^T s_k - \lambda^T s_k^T s_k = u_k \Rightarrow \lambda = \frac{y_k^T s_k - u_k}{s_k^T s_k}$$

$$y_{k+2} = y_k - \frac{s_k (y_k^T s_k - u_k)^T}{s_k^T s_k}$$

$$y_{k+2} u_k = y_k u_k - \frac{s_k (y_k^T s_k - u_k)^T u_k}{s_k^T s_k} = y_k \quad ?$$

$$\left\{ \begin{array}{l} \frac{1}{2} \| y_{k+2} - y_k \|_F^2 \rightarrow \min \\ y_{k+2} \end{array} \right.$$

$$y_{k+2} u_k = y_k \Rightarrow y_{k+2} = y_k - \frac{(y_k u_k - y_k) u_k^T}{u_k^T u_k}$$

$$y_{k+2}^T s_k = y_k^T s_k - \frac{u_k (y_k u_k - y_k)^T s_k}{u_k^T u_k} = u_k$$

$$\left(1 + \frac{(y_k u_k - y_k)^T s_k}{u_k^T u_k} \right) u_k = y_k^T s_k \Rightarrow u_k = \delta_k y_k^T s_k$$

$$\delta_k^2 = \frac{u_k^T s_k}{s_k^T B_k s_k}$$

$$B_{k+1}^{BFGS} = y_{k+2} y_{k+2}^T = \left(y_k - \frac{(y_k u_k - y_k) u_k^T}{u_k^T u_k} \right) \left(y_k - \frac{(y_k u_k - y_k) u_k^T}{u_k^T u_k} \right)^T$$

$$= B_k - \frac{B_k s_k s_k^T B_k}{s_k^T B_k s_k} + \frac{y_k y_k^T}{y_k^T s_k}$$

$$C_k = B_k^{-1}, \quad C_{k+1}^{BFGS} = (I - \rho_k (y_k s_k^T)) C_k (I - \rho_k y_k s_k^T) + \rho_k s_k s_k^T, \quad \rho_k = 1 / y_k^T s_k > 0$$

L - BFGS

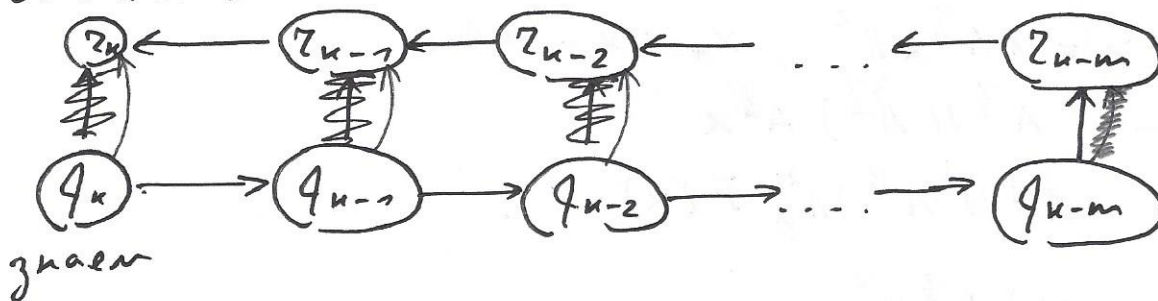
Цель: вычислить $-C_k \nabla f_k$ с помощью $\{y_i, s_i\}_{i=k-m}^k$ через BFGS обновления, начиная

$$C_k^0 = \frac{s_{k-1}^T y_{k-1}}{y_{k-1}^T y_{k-1}} I$$

$$q_k = -\nabla f_k, \quad z_k = C_k q_k = (I - \rho_{k-1}(y_{k-1} s_{k-1}^T)^T) C_{k-1} \cdot$$

$$\begin{aligned} & (I - \rho_{k-1} y_{k-1} s_{k-1}^T) q_k + \rho_{k-1} s_k (s_{k-1}^T q_k) = \\ & = (I - \rho_{k-1}(y_{k-1} s_{k-1}^T)^T) C_{k-1} (\underbrace{q_k - \rho_{k-1} (s_{k-1}^T q_k) y_{k-1}}_{q_{k-1}}) + \\ & + \rho_{k-1} (s_{k-1}^T q_k) s_{k-1} = \\ & = z_{k-1} - \rho_{k-1} \underbrace{y_{k-1}^T z_{k-1}}_{\rho_{k-1}} (s_{k-1}^T z_{k-1}) + \rho_{k-1} (s_{k-1}^T q_k) s_{k-1} \end{aligned}$$

хотим
вычислить



$$z_{k-m} = C_k^0 q_{k-m} = \frac{s_{k-1}^T y_{k-1}}{y_{k-1}^T y_{k-1}} q_{k-m}$$

сложность итерации: $O(m)$, память: $O(m)$

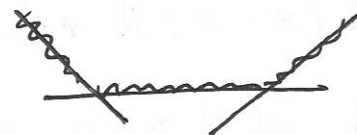
семинар

$$f(x, y) = \begin{cases} \frac{y^2}{x}, & x > 0 \\ 0, & x = y = 0 \end{cases}$$

$$F_t(x, y) = 2ty - x t^2$$

[5]

$$\max_{t \in \mathbb{R}} F_t(x, y) = f(x, y)$$



✓ выпуклая гр-ая представляется как максимум множества линейных функций

$$BFGS \quad x_{k+1}^* = x_k - (\nabla^2 F(x_k))^{-1} \nabla F(x_k)$$

$$\nabla^2 F(x_k) \approx B_k, \quad x_{k+1} = x_k - B_k^{-1} \nabla F(x_k), \quad B_k^{-1} = H_k$$

$$x_{k+1} = x_k - H_k \nabla f(x_k)$$

$$F(x) = \frac{1}{2} \langle Ax, x \rangle, \quad \nabla^2 F(x) = A$$

$$A \in S_{++}^n \quad \text{хотим генерировать} \quad H_k \rightarrow A^{-1}, \quad k \rightarrow \infty$$

$$\textcircled{H} \rightarrow H_+, \quad s \in \mathbb{R}^n, \quad As = y \Leftrightarrow \bar{A}^{-1}y = s \Leftrightarrow \underline{\bar{A}^{-1}As = s}$$

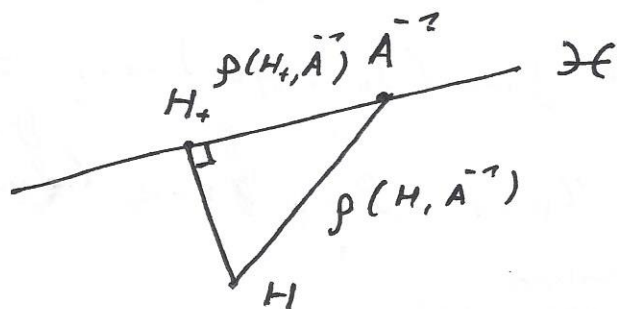
$$\boxed{H_+ As = s} \quad \rho(x, y)$$

$$\mathcal{H} = \{x \in S^n : xAs = s\}$$

$$\min_{x \in S^n} \rho(x, H) : xAs = s$$

$$\rho(x, y) = \|A^{\frac{1}{2}}(x - y)A^{\frac{1}{2}}\|_F$$

$$\rho(H, \bar{A}^{-1}) = \|\bar{A}^{-\frac{1}{2}}HA^{\frac{1}{2}} - I_n\|_F$$



$$F(x) = \langle Ax, x \rangle = \|A^{\frac{1}{2}}x\|^2, \quad x_+ = x - HAx$$

$$A^{\frac{1}{2}}x_+ = A^{\frac{1}{2}}x - (A^{\frac{1}{2}}HA^{\frac{1}{2}})A^{\frac{1}{2}}x$$

$$F(x_+) = \| (I_n - A^{\frac{1}{2}}HA^{\frac{1}{2}}) \|^2_F F(x)$$

$$\begin{cases} \min_{x \in S^{n^2}} \frac{1}{2} \|A^{\frac{1}{2}}(x - H)A^{\frac{1}{2}}\|^2 \\ s.t. \quad xAs = s \end{cases}$$

$$\mathcal{L}(x, \lambda) = \frac{1}{2} \|A^{\frac{1}{2}}(x - H)A^{\frac{1}{2}}\|^2 + \bar{\alpha} \langle \lambda, xAs - s \rangle$$

$$d\mathcal{L}(x, \lambda) = \langle A^{\frac{1}{2}}(x - H)A^{\frac{1}{2}}, A^{\frac{1}{2}}(dx)A^{\frac{1}{2}} \rangle + \bar{\alpha} \langle \lambda, (dx)As \rangle$$

$$= \langle A(x - H)A, dx \rangle + \bar{\alpha} \langle \lambda s^T A, dx \rangle$$

$$\nabla \mathcal{L} = A(x - H)A + \frac{\bar{\alpha}}{2} \lambda s^T A + A s \lambda^T$$

$$\| \langle B, dx \rangle = \langle \frac{1}{2} (B + B^T), dx \rangle$$

$$B \in S^n$$

$$A(X - H)A = \frac{1}{2}(\lambda s^T A + A s \lambda^T)$$

$$X = H + \frac{1}{2} A^{-2}(\lambda s^T A + A s \lambda^T) A^{-2}$$

$$X = H + \frac{1}{2} A^{-2} \lambda s^T + \frac{1}{2} s \lambda^T A^{-2} \rightarrow X A s = s$$

orthogonalisieren s :

$$H A s + \frac{1}{2} A^{-2} \lambda (s^T A s) + \frac{1}{2} s (\lambda^T s) = s \quad \langle A s, s \rangle = 1$$

$$H A s + \frac{1}{2} \langle A s, s \rangle A^{-2} \lambda + \frac{1}{2} \langle \lambda, s \rangle s = s$$

$$H A s + \frac{1}{2} A^{-2} \lambda + \frac{1}{2} \langle \lambda, s \rangle s = s$$

$$A^{-2} \lambda = 2s - \langle \lambda, s \rangle s - 2H A s$$

$$\langle s, \lambda \rangle = \langle s, 2A s - 2A H A s - \langle \lambda, s \rangle A s \rangle$$

$$\langle s, \lambda \rangle = 2 \underbrace{\langle s, A s \rangle}_2 - 2 \langle s, A H A s \rangle - \underbrace{\langle A s, s \rangle}_2 \langle \lambda, s \rangle$$

$$\cancel{\langle \lambda, s \rangle} = 1 - \langle s, A H A s \rangle$$

$$A^{-2} \lambda = 2s - (1 - \langle s, A H A s \rangle) s - 2H A s =$$

$$= s + \langle s, A H A s \rangle s - 2H A s$$

$$X = H + \frac{1}{2} A^{-2} \lambda s^T + \frac{1}{2} s \lambda^T A^{-2}$$

$$X = H + \frac{1}{2} s s^T + \frac{1}{2} \langle s, A H A s \rangle s s^T - H A s s^T +$$

$$+ \frac{1}{2} s s^T + \frac{1}{2} \langle s, A H A s \rangle s s^T - s s^T A H =$$

$$= H + s s^T + \langle s, A H A s \rangle s s^T - H A s s^T - s s^T A H =$$

$$= (I_n - s s^T A) H (I_n - A s s^T)$$

$$H_+ = s s^T + (I_n - s s^T A) H (I - A s s^T)$$

$$H_+ = \frac{s s^T}{\langle A s, s \rangle^2} + \left(I - \frac{s s^T A}{\langle A s, s \rangle^2} \right) H \left(I - \frac{A s s^T}{\langle A s, s \rangle^2} \right)$$

$$y = A s:$$

$$H_+ = \frac{s s^T}{\langle y, s \rangle^2} + \left(I_n - \frac{s y^T}{\langle s, y \rangle^2} \right) H \left(I_n - \frac{y s^T}{\langle s, y \rangle^2} \right)$$

$$F(x) = \frac{1}{2} \langle Ax, x \rangle, \quad y = As = \nabla F(s)$$

$$s = x^1 - x^2, \quad y = Ax^1 - Ax^2 = \nabla F(x^1) - \nabla F(x^2)$$

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

1) если F - строго выпукла, F - непрерывна
 $F(y) \geq F(x) + \langle \nabla F(x), y - x \rangle + \frac{\mu}{2} \|y - x\|^2 \quad \forall x, y \in Q$

2) Q - замкнуто

тогда $\exists!$ минимум

$$\square \quad F(x) \geq F(x^*) + \frac{\mu}{2} \|x - x^*\|^2 \Rightarrow !$$

$\exists ? \quad L_0 = \{x \in Q : F(x) \leq F(x_0)\}$ ограниченное мн-во

$$x \in L_0 \Rightarrow F(x_0) \geq F(x) \geq F(x_0) + \langle \nabla F(x_0), x - x_0 \rangle + \frac{\mu}{2} \|x - x_0\|^2$$

$$\frac{\mu}{2} \|x - x_0\|^2 \leq \langle \nabla F(x_0), x - x_0 \rangle \leq \|\nabla F(x_0)\| \|x - x_0\|$$

$$\|x - x_0\| \leq \frac{2}{\mu} \|\nabla F(x_0)\| \quad \text{мн-во ограничено}$$

минимум \exists . \square

