

① Cauchy Point.

$$m_k(p) = f(x_k) + \nabla f_k^T p + \frac{1}{2} p^T B_k p \Rightarrow \text{Model function}$$

$$\min_{\|p_k\| \leq \Delta_k} m_k(p) \quad \text{Cauchy point}$$

$$\min_{\|tp_k^s\| \leq \Delta_k} m_k(tp_k^s)$$

$$p_k^s = \frac{-\nabla f_k}{\|\nabla f_k\|} \cdot \Delta_k$$

$$\min_{0 \leq t \leq 1} f(x_k) - \|\nabla f_k\| \cdot t \Delta_k + \frac{1}{2} t^2 \left(\frac{\Delta_k}{\|\nabla f_k\|} \right)^2 \cdot \nabla f_k^T B_k \nabla f_k$$

$$\begin{aligned} h'(t) &= -\|\nabla f_k\| \Delta_k + t \left(\frac{\Delta_k}{\|\nabla f_k\|} \right)^2 \cdot \nabla f_k^T B_k \nabla f_k \\ &= 0 \\ \Rightarrow t &= \frac{\|\nabla f_k\| \Delta_k}{\left(\frac{\Delta_k}{\|\nabla f_k\|} \right)^2 \nabla f_k^T B_k \nabla f_k} \end{aligned}$$

$$= \begin{cases} 1, & \nabla f_k^T B_k \nabla f_k < 0 \\ t^*, & \nabla f_k^T B_k \nabla f_k \geq 0 \end{cases}$$

$$= \min \left\{ 1, \frac{\|\nabla f_k\|^3}{\Delta_k \cdot \nabla f_k^T B_k \nabla f_k} \right\}$$

The Diagram of Convergence for Cauchy Point (Global)

重要结论

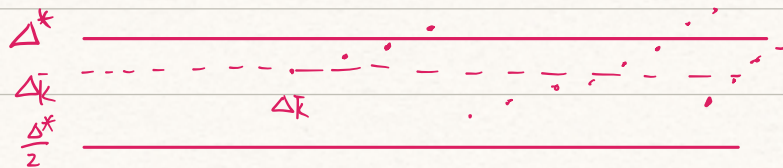
$$\|p_k\| < C \cdot \Delta_k$$



Δ_k 可以 Bound 住 0

$$\|p_k\| \downarrow$$

choose Δ_k 小 可以让 p_k 表现好



永远都会有个下界 Δ



Δ_k 更新策略保证 Δ_k 不能太小

$$\Delta_k \geq \Delta^*$$

结论: Convergence in the sense of $\lim_{n \rightarrow \infty} \|\nabla f_n\| = 0$



Δ_k 小 $\Rightarrow p_k \rightarrow 1 \Rightarrow \Delta_k$ 变大 $\Rightarrow \Delta_k$ bounded from below

i.e., $\Delta_k \geq \Delta$

② { 不能有只有 finite 项不是 $p_k < \frac{1}{4} \Rightarrow \Delta_k \rightarrow 0$
不能存在'无限项' $p_k \geq \frac{1}{4}$

$$\Rightarrow \text{无限项} \Rightarrow \text{充分下降} \Rightarrow \underline{f \rightarrow -\infty}$$

Sufficient & Necessary Condition for TR subprob.

$$p^* = \operatorname{argmin}_p m(p) = f + g^T p + \frac{1}{2} p^T B p \quad \underline{\|p\| \leq \Delta}.$$

$$\Leftrightarrow p^* \text{ is feasible \& } \begin{cases} \exists \lambda \\ (B + \lambda I) p^* = -g \\ \lambda (\|p^*\| - \Delta) = 0 \\ B + \lambda I \succeq 0 \end{cases} \Rightarrow \underline{\text{trivial}} \text{ from KKT condition}$$

$$\Rightarrow \underline{\text{Non-trivial}}$$

→ 理论寻找 p^* :

$$\textcircled{1} B \rightarrow \begin{matrix} \lambda_1, \lambda_2, \dots, \lambda_n \\ \downarrow \qquad \qquad \downarrow \\ q_1 \quad \dots \quad q_n \end{matrix}$$

$$\textcircled{2} \text{ i) } \lambda > -\lambda_1 \Rightarrow B + \lambda I \succ 0 \Rightarrow p^* = -(B + \lambda I)^{-1} g$$

$$= \sum_{i=1}^n - (B + \lambda I)^{-1} \alpha_i q_i \quad \alpha_i = q_i^T g$$

$$= \sum_{i=1}^n - q_i^T g (B + \lambda I)^{-1} q_i$$

$$= \sum_{i=1}^n - q_i^T g \frac{1}{\lambda_i + \lambda} q_i$$

$$= \sum_{i=1}^n - \frac{q_i^T g}{\lambda_i + \lambda} q_i$$

如果 $q_1^T g \neq 0 \Rightarrow$ 寻找 λ s.t. $\|p^*\| = \Delta \Rightarrow \underline{\text{solvable!}}$

如果 $q_1^T g = 0$ 失效

$$\hookrightarrow \text{ii) choose } \lambda = -\lambda_1 \Rightarrow (B + \lambda I) \succeq 0$$

$$\Rightarrow \text{寻找 } (B + \lambda I) p^* = -g$$

$$p^* = \sum \alpha_i q_i \Rightarrow (\lambda_i - \lambda_1) \cdot \alpha_i = -q_i^T g.$$

$$\Rightarrow \alpha_i = \frac{-q_i^T g}{\lambda_i - \lambda_1} \quad \text{for } \underline{i \geq 2}$$

$$\text{when } i=1, \quad \|p^*\| = \left\| \sum_{i=2}^n \alpha_i q_i + \alpha_1 q_1 \right\| = \Delta \Rightarrow \underline{\alpha_1}$$