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# **Trust region methods (L7 and 8): Complete proof of global convergence (optional and non-examinable)**

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C6.2/B2: Continuous Optimization

# Global convergence of the GTR method

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Theorem 11 (GTR global convergence)

Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and bounded below on  $\mathbb{R}^n$ . Let  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1^{(*)}$ . Let  $\{x^k\}$  be generated by the generic trust region (GTR) method, and let the computation of  $s^k$  be such that  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ . Then either

there exists  $k \geq 0$  such that  $\nabla f(x^k) = 0$

or

$$\lim_{k \rightarrow \infty} \|\nabla f(x^k)\| = 0.$$

[ $(*) L \geq 1$  for convenience, to ease calculations.]

We (only) prove  $\liminf_{k \rightarrow \infty} \|\nabla f(x^k)\| = 0$  (which also implies finite termination of GTR) next.

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# Computation of the Cauchy point

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Computation of the Cauchy point: find  $\alpha_c^k$  global solution of

$$\min_{\alpha > 0} m_k(-\alpha \nabla f(x^k)) \text{ subject to } \|\alpha \nabla f(x^k)\| \leq \Delta_k,$$

where  $m_k(s) = f(x_k) + s^T \nabla f(x^k) + \frac{1}{2} s^T \nabla^2 f(x^k) s$ , &  $\nabla f(x^k) \neq 0$ .

■  $\|\alpha \nabla f(x^k)\| \leq \Delta_k \text{ \& } \alpha > 0 \Leftrightarrow 0 < \alpha \leq \frac{\Delta_k}{\|\nabla f(x^k)\|} := \bar{\alpha}.$

■  $\phi(\alpha) := m_k(-\alpha \nabla f(x^k)) = f(x^k) - \alpha \|\nabla f(x^k)\|^2 + \frac{\alpha^2}{2} h^k,$

where  $h^k := \nabla f(x^k)^T \nabla^2 f(x^k) \nabla f(x^k).$

■  $\phi'(0) = -\|\nabla f(x^k)\|^2 < 0$  so  $\phi$  decreasing from  $\alpha = 0$  for suff. small  $\alpha$ ; thus  $\alpha_c^k > 0$ .

■  $h^k > 0$ :  $\alpha_{\min} := \frac{\|\nabla f(x^k)\|^2}{h^k} = \arg \min_{\alpha > 0} \phi(\alpha).$

$\implies \alpha_c^k = \min(\alpha_{\min}, \bar{\alpha}).$

■  $h^k \leq 0$ :  $\phi(\alpha)$  unbounded below on  $\mathbb{R}$  and so  $\alpha_c^k = \bar{\alpha}.$

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# Proof of global convergence of the GTR method

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Lemma 12: (Cauchy model decrease) In GTR with Cauchy decrease  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ , we have the model decrease for each  $k$ ,

$$\begin{aligned} f(x^k) - m_k(s^k) &\geq f(x^k) - m_k(s_c^k) \\ &\geq \frac{1}{2} \|\nabla f(x^k)\| \min \left\{ \Delta_k, \frac{\|\nabla f(x^k)\|}{1 + \|\nabla^2 f(x^k)\|} \right\} \end{aligned}$$

Proof of Lemma 12. (Recall Computation of the Cauchy point)

If  $h^k \leq 0$ , then  $m_k(-\alpha_c^k \nabla f(x^k)) \leq f(x^k) - \alpha_c^k \|\nabla f(x^k)\|^2$ . In this

case, we also have  $\alpha_c^k = \bar{\alpha} = \frac{\Delta_k}{\|\nabla f(x^k)\|}$  and so

$$f(x^k) - m_k(s_c^k) \geq \Delta_k \|\nabla f(x^k)\|.$$

Else,  $h^k > 0$ ; then  $\alpha_c^k = \min\{\alpha_{\min}, \bar{\alpha}\}$  where  $\alpha_{\min} = \|\nabla f(x^k)\|^2 / h^k$ .

Assume first that  $\alpha_c^k = \bar{\alpha}$ . Then  $\alpha_c^k h^k \leq \|\nabla f(x^k)\|^2$  and

$$f(x^k) - m_k(s_c^k) = \alpha_c^k \|\nabla f(x^k)\|^2 - \frac{(\alpha_c^k)^2}{2} h^k \geq \frac{\alpha_c^k}{2} \|\nabla f(x^k)\|^2,$$

# Proof of global convergence of the GTR method

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## Proof of Lemma 12 (continued).

and using the expression of  $\bar{\alpha}$ ,

$$f(x^k) - m_k(s_c^k) \geq \frac{\Delta_k}{2\|\nabla f(x^k)\|} \|\nabla f(x^k)\|^2 = \frac{1}{2} \Delta_k \|\nabla f(x^k)\|.$$

Finally, let  $\alpha_c^k = \alpha_{\min} = \|\nabla f(x^k)\|^2 / h^k$ . Replacing this value in the model decrease we get

$$f(x^k) - m_k(s_c^k) = \alpha_c^k \|\nabla f(x^k)\|^2 - \frac{(\alpha_c^k)^2}{2} h^k = \frac{\|\nabla f(x^k)\|^4}{2h^k},$$

and further, by Cauchy-Schwarz and Rayleigh quotient inequalities,

$$\begin{aligned} \frac{\|\nabla f(x^k)\|^4}{2h^k} &= \frac{\|\nabla f(x^k)\|^4}{2(\nabla f(x^k))^T \nabla^2 f(x^k) \nabla f(x^k)} \\ &\geq \frac{\|\nabla f(x^k)\|^2}{2\|\nabla^2 f(x^k)\|} \geq \frac{\|\nabla f(x^k)\|^2}{2(1+\|\nabla^2 f(x^k)\|)} \quad (*) \end{aligned}$$

Thus  $f(x^k) - m_k(s_c^k) \geq \frac{\|\nabla f(x^k)\|^2}{2(1+\|\nabla^2 f(x^k)\|)}$ .  $\square$

[(\*) '+1' is only needed to cover the case  $H^k = 0$ .]

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**Lemma 13: (Model error bound)** Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L$ . Then in GTR, for all  $k \geq 0$ , we have  $|f(x^k + s^k) - m_k(s^k)| \leq L\Delta_k^2$ .

**Proof of Lemma 13.** Mean-value theorem gives

$$f(x^k + s^k) = f(x^k) + (s^k)^T \nabla f(x^k) + \frac{1}{2}(s^k)^T \nabla^2 f(\xi^k) s^k$$

for some  $\xi^k$  on line segment  $[x^k, x^k + s^k]$ . Then the definition of  $m_k(s) = f(x^k) + s^T \nabla f(x^k) + \frac{1}{2}s^T \nabla^2 f(x^k) s$  gives

$$\begin{aligned} |f(x^k + s^k) - m_k(s^k)| &\leq \frac{1}{2} |(s^k)^T \nabla^2 f(\xi^k) s^k - (s^k)^T \nabla^2 f(x^k) s^k| \\ &\leq \frac{1}{2} |(s^k)^T \nabla^2 f(\xi^k) s^k| + \frac{1}{2} |(s^k)^T \nabla^2 f(x^k) s^k| \\ &\leq \frac{1}{2} [\|\nabla^2 f(\xi^k)\| + \|\nabla^2 f(x^k)\|] \cdot \|s^k\|^2 \leq L\|s^k\|^2 \leq L\Delta_k^2, \end{aligned}$$

where in the penultimate inequality we used that ( $\nabla f$  Lipschitz continuous with const.  $L$ )  $\iff$  ( $\|\nabla^2 f\|$  uniformly bounded above by  $L$ ), and in the last inequality we used that  $\|s^k\| \leq \Delta_k$ .  $\square$

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# Proof of global convergence of the GTR method

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**Lemma 14: (Successful iterations)** Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1$ . In GTR with Cauchy decrease  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ , suppose that  $\nabla f(x^k) \neq 0$  and

$$\Delta_k \leq \frac{0.45}{L} \|\nabla f(x^k)\|. \quad (1)$$

Then iteration  $k$  is successful and  $\Delta_{k+1} \geq \Delta_k$ .

**Proof of Lemma 14.**  $\nabla f$  Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1 \implies 1 + \|\nabla^2 f(x)\| \leq 2L$  and so from

(1), we deduce  $\Delta_k \leq \frac{\|\nabla f(x^k)\|}{1 + \|\nabla^2 f(x^k)\|}$ . Lemma 12 now gives that

$$f(x^k) - m_k(s^k) \geq \frac{1}{2} \|\nabla f(x^k)\| \Delta_k > 0$$

and Lemma 13 that  $|f(x^k + s^k) - m_k(s^k)| \leq L\Delta_k^2$ . We evaluate

$$|\rho_k - 1| = \frac{|f(x^k + s^k) - m_k(s^k)|}{f(x^k) - m_k(s^k)} \leq \frac{2L\Delta_k^2}{\Delta_k \|\nabla f(x^k)\|} = \frac{2L\Delta_k}{\|\nabla f(x^k)\|} \leq 0.9 \implies \rho_k \geq 0.1. \quad \square$$

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**Lemma 15:** (Lower bound on TR radius) Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1$ . In GTR with Cauchy decrease  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ , suppose that there exists  $\epsilon > 0$  such that  $\|\nabla f(x^k)\| \geq \epsilon$  for all  $k$ . Then

$$\Delta_k \geq \frac{0.45}{2L} \epsilon \quad \text{for all } k \geq 0.$$

**Proof of Lemma 15.** Assume the contrary:  $k$  is the first iteration such that  $\Delta_{k+1} < \frac{0.45}{2L} \epsilon$ . Then  $k$  unsuccessful and  $\Delta_{k+1} = \frac{\Delta_k}{2}$ . Thus  $\Delta_k = 2\Delta_{k+1} < \frac{0.45}{L} \epsilon \leq \frac{0.45}{L} \|\nabla f(x^k)\|$  and so by Lemma 14,  $k$  must be successful, contradiction.  $\square$



# Proof of global convergence of the GTR method

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Theorem 16: (The case of finitely many successful iterations)

Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1$ . In GTR with Cauchy decrease  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ , suppose that there are finitely many successful iterations that occur. Then  $x^k = x_*$  for all  $k$  sufficiently large and  $\nabla f(x_*) = 0$ .

Proof of Theorem 16. Let  $k_o$  be the last successful iteration. Then GTR implies  $x^k = x_*$  for all  $k \geq k_o + 1$ . As all remaining iterations are unsuccessful,  $\Delta_{k+1} = \frac{1}{2}\Delta_k$  for all  $k \geq k_o + 1$  and so  $\Delta_k \rightarrow 0$  as  $k \rightarrow \infty$ . If  $\nabla f(x^{k_o+1}) \neq 0$ , then let  $\epsilon = \|\nabla f(x^{k_o+1})\|$  in Lemma 15, which implies that  $\Delta_k$  is bounded away from zero; contradiction. Thus  $\nabla f(x^{k_o+1}) = 0$  and so  $\nabla f(x^k) = \nabla f(x_*) = 0$  for all  $k \geq k_o + 1$ .  $\square$

# Proof of global convergence of the GTR method

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Theorem 17: (At least one limit point is stationary) Let  $f \in \mathcal{C}^2(\mathbb{R}^n)$  and and bounded below on  $\mathbb{R}^n$ . Let  $\nabla f$  be Lipschitz continuous on  $\mathbb{R}^n$  with Lipschitz constant  $L \geq 1$ . Let  $\{x^k\}$  be generated by the generic trust region (GTR) method, and let the computation of  $s^k$  be such that  $m_k(s^k) \leq m_k(s_c^k)$  for all  $k$ . Then either there exists  $k \geq 0$  such that  $\nabla f(x^k) = 0$  or  $\liminf_{k \rightarrow \infty} \|\nabla f(x^k)\| = 0$ .

Proof of Theorem 17. If there exists  $k$  such that  $\nabla f(x^k) = 0$ , then GTR terminates. Assume there exists  $\epsilon > 0$  such that  $\|\nabla f(x^k)\| \geq \epsilon$  for all  $k$ . Then Th 16 implies that there are infinitely many successful iterations  $k \in \mathcal{S}$ , and from GTR/ $\rho_k$ ,

$$\begin{aligned} f(x^k) - f(x^{k+1}) &\geq 0.1(f(x^k) - m_k(s^k)) \\ &\geq \frac{0.1}{2} \|\nabla f(x^k)\| \min\left\{\frac{\|\nabla f(x^k)\|}{1 + \|\nabla^2 f(x^k)\|}, \Delta_k\right\} \end{aligned}$$

for all  $k \in \mathcal{S}$ , where we also used Lemma 12.

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# Proof of global convergence of the GTR method

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## Proof of Theorem 17 (continued).

$\nabla f$  Lipschitz cont. with Lips const  $L \geq 1 \implies \|\nabla^2 f(x)\| \leq L \forall x$ .

Thus since  $\|\nabla f(x^k)\| \geq \epsilon$  for all  $k$ , we have for all  $k \in \mathcal{S}$  that

$$f(x^k) - f(x^{k+1}) \geq 0.05\epsilon \min \left\{ \frac{\epsilon}{2L}, \Delta_k \right\} \geq 0.05\epsilon \min \left\{ \frac{\epsilon}{2L}, \frac{0.45}{2L}\epsilon \right\},$$

where we also used Lemma 15. Thus

$$\text{for all } k \in \mathcal{S}: f(x^k) - f(x^{k+1}) \geq \frac{0.01}{2L}\epsilon^2. \quad (*)$$

Since  $f(x^k) \geq f_{\text{low}}$  for all  $k$ , we deduce

$$\begin{aligned} f(x^0) - f_{\text{low}} &\geq f(x^0) - \lim_{k \rightarrow \infty} f(x^k) \geq \sum_{i=0}^{\infty} (f(x^i) - f(x^{i+1})) \\ &= \sum_{i \in \mathcal{S}} (f(x^i) - f(x^{i+1})) \geq |\mathcal{S}| \frac{0.01}{2L}\epsilon^2 \quad (**) \end{aligned}$$

where in '=' we used  $f(x^k) = f(x^{k+1})$  on all unsuccessful  $k$ , and in the last ' $\geq$ ', we used (\*) and  $|\mathcal{S}| = \text{no. of successful iterations}$ . But LHS of (\*\*) is finite while RHS of (\*\*) is infinite since  $|\mathcal{S}| = \infty$ . Thus there must exist  $k$  such that  $\|\nabla f(x^k)\| < \epsilon$ .  $\square$

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