MINIMIZATION OF FUNCTIONS HAVING LIPSCHITZ CONTINUOUS FIRST PARTIAL DERIVATIVES

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A general convergence theorem for the gradient method is proved under hypotheses which are given below. It is then shown that the usual steepest descent and modified steepest descent algorithms converge under the some hypotheses. The modified steepest descent algorithm allows for the possibility of variable stepsize.

For a comparison of our results with results previously obtained, the reader is referred to the discussion at the end of this paper.

Principal conditions. Let f be a real-valued function defined and continuous everywhere on E^n (real Euclidean *n*-space) and bounded below E^n . For fixed $x_0 \in E^n$ define $S(x_0) = \{x : f(x) \leq f(x_0)\}$. The function f satisfies: condition I if there exists a unique point $x^* \in E^n$ such that $f(x^*) = \inf_{x \in E^n} f(x)$; Condition II at x_0 if $f \in C^1$ on $S(x_0)$ and $\nabla f(x) = 0$ for $x \in S(x_0)$ if and only if $x = x^*$; Condition III at x_0 if $f \in C^1$ on $S(x_0)$ and ∇f is Lipschitz continuous on $S(x_0)$, i.e., there exists a Lipschitz constant K > 0 such that $|\nabla f(y) - \nabla f(x)| \leq K |y - x|$ for every pair $x, y \in S(x_0)$; Condition IV at x_0 if $f \in C^1$ on $S(x_0)$ and if r > 0 implies that m(r) > 0 where $m(r) = \inf_{x \in S_r(x_0)} |\nabla f(x)|$, $S_r(x_0) = S_r \cap S(x_0)$, $S_r =$ $\{x : |x - x^*| \geq r\}$, and x^* is any point for which $f(x^*) = \inf_{x \in E^n} f(x)$. (If $S_r(x_0)$ is void, we define $m(r) = \infty$.)

It follows immediately from the definitions of Conditions I through IV that Condition IV implies Conditions I and II, and if $S(x_0)$ is bounded, then Condition IV is equivalent to Conditions I and II.

2. The convergence theorem. In the convergence theorem and its corollaries, we will assume that f is a real-valued function defined and continuous everywhere on E^n , bounded below on E^n , and that Conditions III and IV hold at x_0 .

THEOREM. If $0 < \delta \leq 1/4K$, then for any $x \in S(x_0)$, the set (1) $S^*(x, \delta) = \{x_{\lambda}: x_{\lambda} = x - \lambda \nabla f(x), \ \lambda > 0, \ f(x_{\lambda}) - f(x) \leq -\delta \ |\nabla f(x)|^2 \}$

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is a nonempty subset of $S(x_0)$ and any sequence $\{x_k\}_{k=0}^{\infty}$ such that $x_{k+1} \in S^*(x_k, \delta)$, $k = 0, 1, 2, \cdots$, converges to the point x^* which minimizes f.

Proof. If $x \in S(x_0)$, $x_{\lambda} = x - \lambda \nabla f(x)$ and $0 \leq \lambda \leq 1/K$, Condition III and the mean value theorem imply the inequality $f(x_{\lambda}) - f(x) \leq -(\lambda - \lambda^2 K) |\nabla f(x)|^2$ which in turn implies that $x_{\lambda} \in S^*(x, \delta)$ for

$$\lambda_1 \leqq \lambda \leqq \lambda_2$$
 , $\lambda_i = rac{1}{2K} [1 + (-1)^i \sqrt{1 - 4 \delta K}]$,

so that $S^*(x, \delta)$ is a nonempty subset of $S(x_0)$. If $\{x_k\}_{k=0}^{\infty}$ is any sequence for which $x_{k+1} \in S^*(x_k, \delta)$, $k = 0, 1, 2, \cdots$, then (1) implies that sequence $\{f(x_k)\}_{k=0}^{\infty}$, which is bounded below, is monotone nonincreasing and hence that $|\nabla f(x_k)| \to 0$ as $k \to \infty$. The remainder of the theorem follows from Condition IV.

COROLLARY 1. (The Steepest Descent Algorithm) If

$$x_{\scriptscriptstyle k+1} = x_{\scriptscriptstyle k} - rac{1}{2K}
abla f(x_{\scriptscriptstyle k}), \,\, k=0,\,1,\,2,\,\cdots$$

then the sequence $\{x_k\}_{k=0}^{\infty}$ converges to the point x^* which minimizes f.

Proof. It follows from the proof of the convergence theorem that the sequence $\{x_k\}_{k=0}^{\infty}$ defined in the statement of Corollary 1 is such that $x_{k+1} \in S^*(x_k, 1/4K), k = 0, 1, 2, \cdots$.

COROLLARY 2. (The Modified Steepest Descent Algorithm) If α is an arbitrarily assigned positive number, $\alpha_m = \alpha/2^{m-1}$, $m = 1, 2, \cdots$, and $x_{k+1} = x_k - \alpha_{m_k} \nabla f(x_k)$ where m_k is the smallest positive integer for which

$$(2) \qquad f(x_k - lpha_{m_k} \nabla f(x_k)) - f(x_k) \leq - rac{1}{2} lpha_{m_k} | \nabla f(x_k) |^2 \, ,$$

 $k = 0, 1, 2, \cdots$, then the sequence $\{x_k\}_{k=0}^{\infty}$ converges to the point x^* which minimizes f.

Proof. It follows from the proof of the convergence theorem that if $x \in S(x_0)$ and $x_{\lambda} = x - \lambda \nabla f(x)$, then $f(x_{\lambda}) - f(x) \leq -(1/2)\lambda |\nabla f(x)|^2$ for $0 \leq \lambda \leq 1/2K$. If $\alpha \leq 1/2K$, then for the sequence $\{x_k\}_{k=0}^{\infty}$ in the statement of Corollary 2, $m_k = 1$ and $x_{k+1} \in S^*(x_k, (1/2)\alpha), k = 0, 1, 2, \cdots$. If $\alpha > 1/2K$, then the integers m_k exist and $\alpha_{m_k} > 1/4K$ so that $x_{k+1} \in S^*(x_k, 1/8K), k = 0, 1, 2, \cdots$. 3. Discussion. The convergence theorem proves convergence under hypotheses which are more restrictive than those imposed by Curry [1] but less restrictive than those imposed by Goldstein [2]. However, both the algorithms which we have considered would be considerably easier to apply than the algorithm proposed by Curry since his algorithm requires the minimization of a function of one variable at each step. The method of Goldstein requires the assumption that $f \in C^2$ on $S(x_0)$ and that $S(x_0)$ be bounded. It also requires knowledge of a bound for the norm of the Hessian matrix of f on $S(x_0)$, but yields an estimate for the ultimate rate of convergence of the gradient method. It should be pointed out that the modified steepest descent algorithm of Corollary 2 allows for the possibility of variable stepsize and does not require knowledge of the value of the Lipschitz constant K.

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References

1. H. B. Curry, The method of steepest descent for nonlinear minimization problems, Quart. Appl. Math. 2 (1944), 258-263.

2. A. A. Goldstein, Cauchy's method of minimization, Numer. Math. 4 (2), (1962), 146-150.