

Math 408A

Bisection Method for the Weak Wolfe Conditions

Wolfe Conditions

A Bisection Method for the Weak Wolfe Conditions

Weak Wolf Decsent Algorithm

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Strong Wolfe Conditions:

$$\begin{aligned} f(x^k + t_k d^k) &\leq f(x^k) + c_1 t_k f'(x^k; d^k) \\ |f'(x^k + t_k d^k; d^k)| &\leq c_2 |f'(x^k; d^k)| . \end{aligned}$$

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Weak Wolfe Conditions:

$$\begin{aligned} f(x^k + t_k d^k) &\leq f(x^k) + c_1 t_k f'(x^k; d^k) \\ c_2 f'(x^k; d^k) &\leq f'(x^k + t_k d^k; d^k) . \end{aligned}$$

Wolfe Conditions

Lemma: Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be continuously differentiable and suppose that $x, d \in \mathbb{R}^n$ are such that the set $\{f(x + td) : t \geq 0\}$ is bounded below and $f'(x; d) < 0$, then for each $0 < c_1 < c_2 < 1$ the set

$$\left\{ t \mid \begin{array}{l} t > 0, f'(x + td; d) \geq c_2 f'(x; d), \text{ and} \\ f(x + td) \leq f(x) + c_1 t f'(x; d) \end{array} \right\}$$

has non-empty interior.

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Set $\phi(t) = f(x + td) - (f(x) + c_1 tf'(x; d))$. Then

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MVT $\implies \exists \tilde{t} \in (0, t^*)$ with $\phi'(\tilde{t}) = 0$. That is,

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$$\nabla f(x + \tilde{t}d)^T d = c_1 \nabla f(x)^T d > c_2 \nabla f(x)^T d.$$

We also have

$$f(x + td) - (f(x) + c_1 \tilde{t} \nabla f(x)^T d) < 0.$$

A Bisection Method for the Weak Wolfe Conditions

INITIALIZATION: Choose $0 < c_1 < c_2 < 1$, and set $\alpha = 0$, $t = 1$, and $\beta = +\infty$.

REPEAT

If $f(x + td) > f(x) + c_1 tf'(x; d)$,
set $\beta = t$ and reset $t = \frac{1}{2}(\alpha + \beta)$.

Else if $f'(x + td; d) < c_2 f'(x; d)$,
set $\alpha = t$ and reset

$$t = \begin{cases} 2\alpha, & \text{if } \beta = +\infty \\ \frac{1}{2}(\alpha + \beta), & \text{otherwise.} \end{cases}$$

Else, STOP.

END REPEAT

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Step 3: Set $x^{k+1} = x^k + t_k d^k$ and reset $k = k + 1$.

Return to Step 1.

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- (i) The algorithm terminates finitely at a 1st-order stationary point for f .
- (ii) For some k the stepsize selection procedure generates a sequence of trial stepsizes $t_{k\nu} \uparrow +\infty$ such that $f(x^k + t_{k\nu}d^k) \rightarrow -\infty$.

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- (iii) $f(x^k) \downarrow -\infty$.
- (iv) $\sum_{k=0}^{\infty} \|\nabla f(x^k)\|^2 \cos^2 \theta_k < +\infty$, where $\cos \theta_k = \frac{\nabla f(x^k)^T d^k}{\|\nabla f(x^k)\| \|d^k\|}$ for all $k = 1, 2, \dots$.

Weak Wolf Convergence: Corollary

Let f and $\{x^k\}$ be as in the Theorem, and let $\{H_k\}$ be a sequence of symmetric positive definite matrices for which there exists $\bar{\lambda} > \underline{\lambda} > 0$ such that

$$\underline{\lambda}\|u\|^2 \leq u^T H_k u \leq \bar{\lambda}\|u\|^2 \quad \forall u \in \mathbb{R}^n \text{ and } k = 1, 2, \dots$$

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If the search directions d^k are given by

$$d^k = -H_k \nabla f(x^k) \quad \forall k = 1, 2, \dots,$$

then $\nabla f(x^k) \rightarrow 0$.