It can be shown that if $x_k \to x^*$ superlinearly, then the ratio in this expression converges to 1. If we adjust the choice (3.60) by setting

$$\alpha_0 \leftarrow \min(1, 1.01\alpha_0),$$

we find that the unit step length $\alpha_0 = 1$ will eventually always be tried and accepted, and the superlinear convergence properties of Newton and quasi-Newton methods will be observed.

A LINE SEARCH ALGORITHM FOR THE WOLFE CONDITIONS

The Wolfe (or strong Wolfe) conditions are among the most widely applicable and useful termination conditions. We now describe in some detail a one-dimensional search procedure that is guaranteed to find a step length satisfying the *strong* Wolfe conditions (3.7) for any parameters c_1 and c_2 satisfying $0 < c_1 < c_2 < 1$. As before, we assume that p is a descent direction and that f is bounded below along the direction p.

The algorithm has two stages. This first stage begins with a trial estimate α_1 , and keeps increasing it until it finds either an acceptable step length or an interval that brackets the desired step lengths. In the latter case, the second stage is invoked by calling a function called **zoom** (Algorithm 3.6, below), which successively decreases the size of the interval until an acceptable step length is identified.

A formal specification of the line search algorithm follows. We refer to (3.7a) as the *sufficient decrease condition* and to (3.7b) as the *curvature condition*. The parameter α_{max} is a user-supplied bound on the maximum step length allowed. The line search algorithm terminates with α_* set to a step length that satisfies the strong Wolfe conditions.

```
Set \alpha_0 \leftarrow 0, choose \alpha_{\max} > 0 and \alpha_1 \in (0, \alpha_{\max}); i \leftarrow 1; repeat

Evaluate \phi(\alpha_i); if \phi(\alpha_i) > \phi(0) + c_1\alpha_i\phi'(0) or [\phi(\alpha_i) \ge \phi(\alpha_{i-1}) \text{ and } i > 1]

\alpha_* \leftarrow \mathbf{zoom}(\alpha_{i-1}, \alpha_i) and \mathbf{stop};

Evaluate \phi'(\alpha_i);

if |\phi'(\alpha_i)| \le -c_2\phi'(0)

set \alpha_* \leftarrow \alpha_i and \mathbf{stop};

if \phi'(\alpha_i) \ge 0

set \alpha_* \leftarrow \mathbf{zoom}(\alpha_i, \alpha_{i-1}) and \mathbf{stop};

Choose \alpha_{i+1} \in (\alpha_i, \alpha_{\max});

i \leftarrow i + 1;
```

Algorithm 3.5 (Line Search Algorithm).

end (repeat)

Note that the sequence of trial step lengths $\{\alpha_i\}$ is monotonically increasing, but that the order of the arguments supplied to the **zoom** function may vary. The procedure uses the knowledge that the interval (α_{i-1}, α_i) contains step lengths satisfying the strong Wolfe conditions if one of the following three conditions is satisfied:

(i) α_i violates the sufficient decrease condition;

```
(ii) \phi(\alpha_i) \geq \phi(\alpha_{i-1});
```

(iii)
$$\phi'(\alpha_i) \geq 0$$
.

The last step of the algorithm performs extrapolation to find the next trial value α_{i+1} . To implement this step we can use approaches like the interpolation procedures above, or we can simply set α_{i+1} to some constant multiple of α_i . Whichever strategy we use, it is important that the successive steps increase quickly enough to reach the upper limit α_{max} in a finite number of iterations.

We now specify the function **zoom**, which requires a little explanation. The order of its input arguments is such that each call has the form **zoom**(α_{lo} , α_{hi}), where

- (a) the interval bounded by α_{lo} and α_{hi} contains step lengths that satisfy the strong Wolfe conditions;
- (b) α_{lo} is, among all step lengths generated so far and satisfying the sufficient decrease condition, the one giving the smallest function value; and
- (c) α_{hi} is chosen so that $\phi'(\alpha_{lo})(\alpha_{hi} \alpha_{lo}) < 0$.

Each iteration of **zoom** generates an iterate α_j between α_{lo} and α_{hi} , and then replaces one of these endpoints by α_j in such a way that the properties (a), (b), and (c) continue to hold.

```
Algorithm 3.6 (zoom). repeat  \begin{array}{l} \text{Interpolate (using quadratic, cubic, or bisection) to find} \\ \text{a trial step length } \alpha_j \text{ between } \alpha_{\text{lo}} \text{ and } \alpha_{\text{hi}}; \\ \text{Evaluate } \phi(\alpha_j); \\ \text{if } \phi(\alpha_j) > \phi(0) + c_1\alpha_j\phi'(0) \text{ or } \phi(\alpha_j) \geq \phi(\alpha_{\text{lo}}) \\ \alpha_{\text{hi}} \leftarrow \alpha_j; \\ \text{else} \\ \text{Evaluate } \phi'(\alpha_j); \\ \text{if } |\phi'(\alpha_j)| \leq -c_2\phi'(0) \\ \text{Set } \alpha_* \leftarrow \alpha_j \text{ and stop}; \\ \text{if } \phi'(\alpha_j)(\alpha_{\text{hi}} - \alpha_{\text{lo}}) \geq 0 \\ \alpha_{\text{hi}} \leftarrow \alpha_{\text{lo}}; \\ \alpha_{\text{lo}} \leftarrow \alpha_j; \\ \text{end (repeat)} \end{array}
```