One-dimensional Search Methods

Chapter 7

Newton's Method



• By Taylor's expansion

$$g(x) = f(x^{(k)}) + f'(x^{(k)})(x - x^{(k)}) + \frac{1}{2}f''(x^{(k)})(x - x^{(k)})^2$$

• By the first-order necessary condition

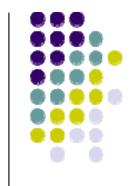
$$g'(x) = 0 = f'(x^{(k)}) + f''(x^{(k)})(x - x^{(k)}) \quad \text{extreme}$$

$$\therefore x^{(k+1)} - x - x^{(k)} - \frac{f'(x^{(k)})}{f''(x^{(k)})}$$

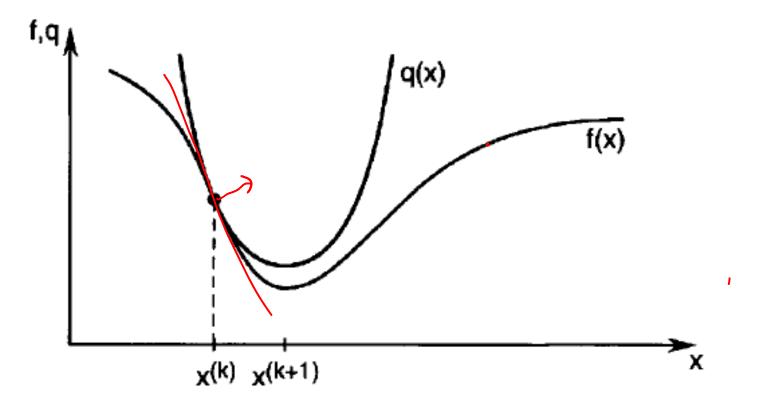
• Example: Find the minimizer of

$$f(x) = \frac{1}{2}x^2 - \cos x$$

Newton's Method (Cont.)



 Note: Newton's method works well when f''(x)>0 everywhere (see Figure 7.6).



Newton's Method (Cont.)



 However, it f''(x)<0 for some x, Newton's method may fail to converge to the minimizer (see Figure 7.7).

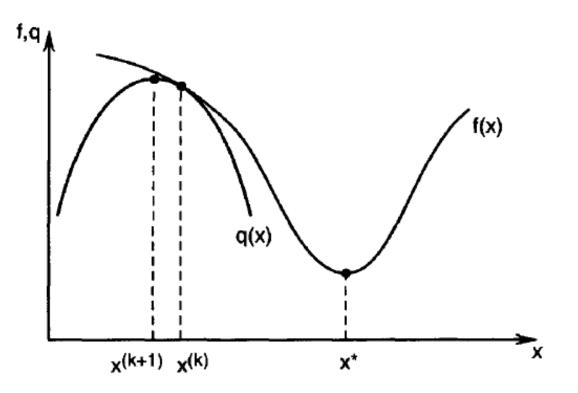
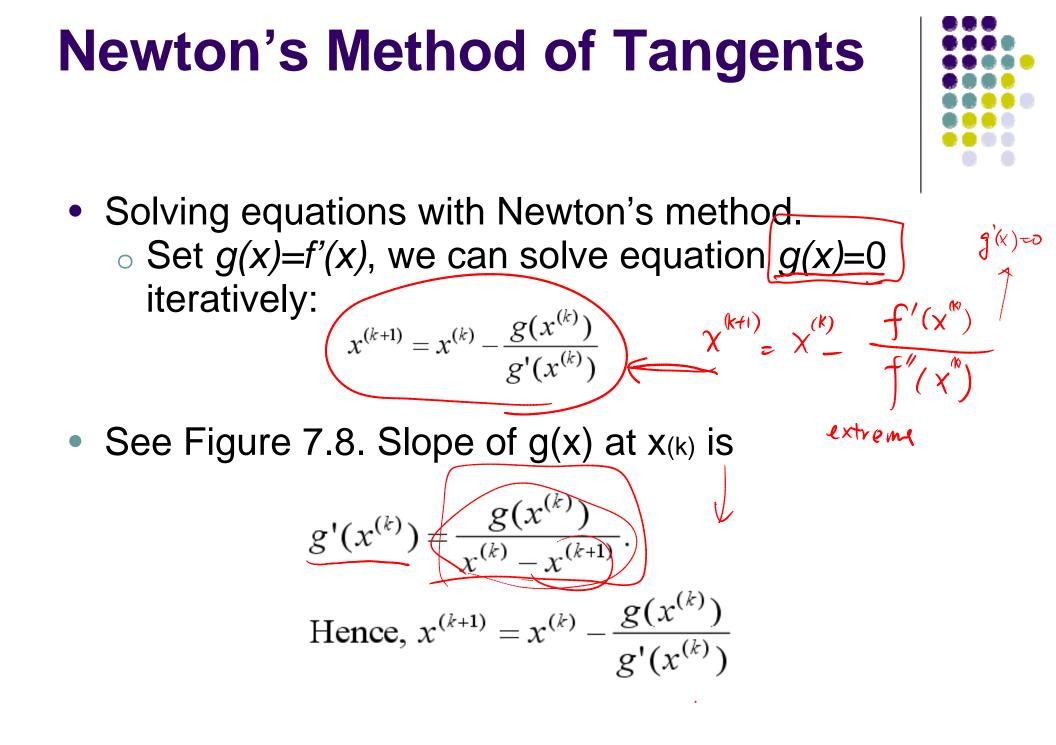


Figure 7.7 Newton's algorithm with f''(x) < 0



Newton's Method of Tangents (cont.) g(x) g(x) 2'(x*) g(x^(k)

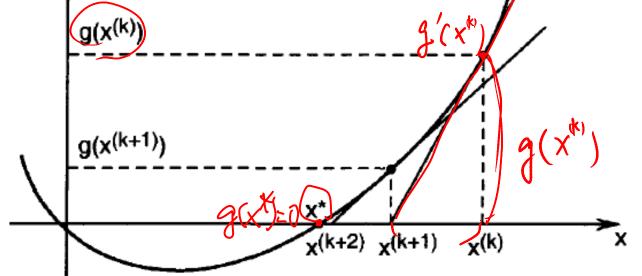


Figure 7.8 Newton's method of tangents

Example:

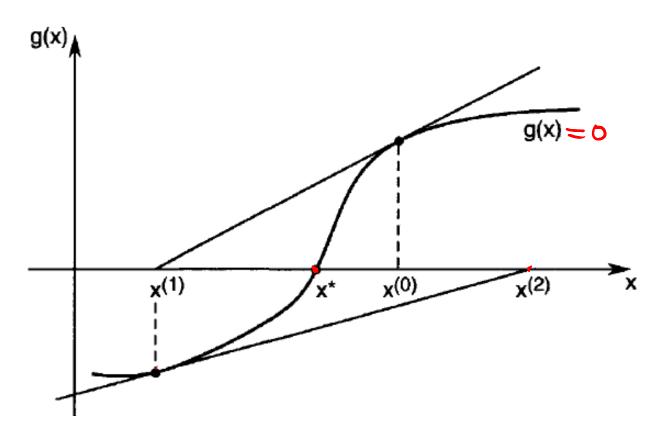


Example 7.4 We apply Newton's method to improve a first approximation, $x^{(0)} = 12$, to the root of the equation

$$g(x) = x^3 - 12.2x^2 + 7.45x + 42 = 0.$$

Newton's Method of Tangents (cont.)

Note: Newton's method of tangents may fail if the first approximation to the root is such that the ratio $g(x_{(0)})/g'(x_{(0)})$ is not small enough (see Figure 7.9).



SECANT Method



• Approximate 2nd-order derivative of f(x) by

$$f''(x^{(k)}) = \frac{f'(x^{(k)}) - f'(x^{(k-1)})}{x^{(k)} - x^{(k-1)}}.$$

Thus,
Equivalently,
See Figure 7.10
$$x^{(k+1)} = \frac{x^{(k)} - \frac{x^{(k)} - x^{(k-1)}}{f'(x^{(k)}) - f'(x^{(k-1)})}f'(x^{(k)}).$$

$$x^{(k+1)} = \frac{x^{(k-1)}f'(x^{(k)}) - x^{(k)}f'(x^{(k-1)})}{f'(x^{(k)}) - f'(x^{(k-1)})}.$$

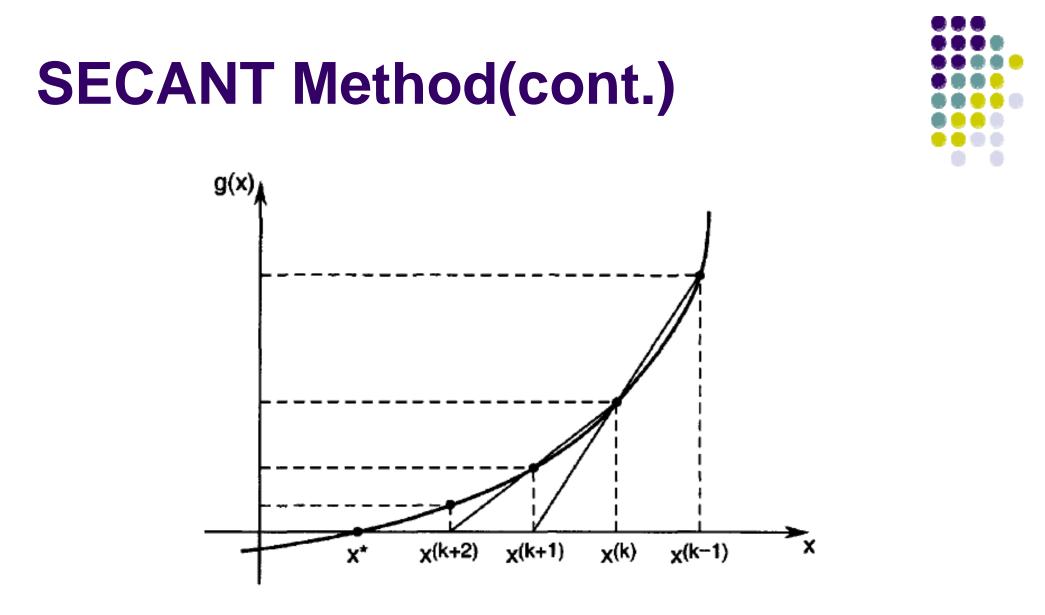


Figure 7.10 Secant method for root finding

Example:



Example 7.5 We apply the secant method to find the root of the equation

$$g(x) = x^3 - 12.2x^2 + 7.45x + 42 = 0.$$

Example 7.6 Suppose the voltage across a resistor in a circuit decays according to the model $V(t) = e^{-Rt}$, where V(t) is the voltage at time t, and R is the resistance value.

Remarks



- Iterative algorithms for multidimensional optimization problem typically involve a <u>line search</u> at every iteration. $\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} + \alpha_{\scriptscriptstyle L} \mathbf{d}^{(k)}$
- Search direction: *d*_(k)
 α_(k) ≥0 is chosen to minimize φ_k(α) = f(x^(k) + αd^(k))
 Using secant method to find minimal φ_k(α) needs the

derivative of ϕ_k which is

$$\phi'_{k}(\alpha) = \mathbf{d}^{(k)^{T}} \nabla f(\mathbf{x}^{(k)} + \alpha \mathbf{d}^{(k)}).$$

