MA 204 Numerical Methods

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January 11, 2024

Contents

 Solution of a nonlinear equation, bisection and secant methods, Newton's method, rate of convergence, solution of a system of nonlinear equations.

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- Interpolation by polynomials, divided differences, error of the interpolating polynomial, piecewise linear and cubic spline interpolation.

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■ The iterative procedure (given below) is called the **Newton** Raphson's method.

$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}, \ n = 0, 1, 2, \cdots.$$

History

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- The most basic version starts with a real-valued function f, its derivative f, and an initial guess x_0 for a root of f.
- This is one of the most powerful methods for solving a root-finding problem. There are many ways of introducing Newton's method.

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 $(x_1,0)$ lies on this line:

$$\implies 0 = f(x_0) + f'(x_0)(x_1 - x_0)$$

$$\implies x_1 = x_0 - \frac{f(x_0)}{f'(x_0)}.$$

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Repeating this process, we obtain a sequence of numbers x_1, x_2, \cdots that we hope will approach the root α . These numbers are called iterates, and they are defined recursively by the following general iteration formula:

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$$x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}, \quad n = 0, 1, 2, \cdots$$
 (1)

This is **Newton's method** for solving f(x) = 0.

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- This says that the graph of y = f(x) is not tangent to the x-axis when the graph intersects it $x = \alpha$. This case when $f'(\alpha) = 0$ will be discussed later.
- Combining (2) with the continuity of f'(x) implies

$$f'(x) \neq 0 \quad \forall \ x \text{ near } \alpha.$$

By Taylor's theorem,

$$f(\alpha) = f(x_n + \alpha - x_n)$$

$$= f(x_n) + (\alpha - x_n)f'(x_n) + \frac{(\alpha - x_n)^2}{2}f''(c_n)$$
 (3)

where c_n is an unknown point between α and x_n .

Note that $f(\alpha) = 0$. Using $x_{n+1} = x_n - \frac{f(x_n)}{f'(x_n)}$ This implies

$$0 = f(x_n) + (\alpha - x_n)f'(x_n) + \frac{1}{2}(\alpha - x_n)^2 f''(c_n)$$

$$\implies 0 = \frac{f(x_n)}{f'(x_n)} + (\alpha - x_n) + \frac{1}{2}(\alpha - x_n)^2 \frac{f''(c_n)}{f'(x_n)}$$

$$\implies 0 = x_n - x_{n+1} + \alpha - x_n + (\alpha - x_n)^2 \frac{f''(c_n)}{f'(x_n)}$$
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Error Analysis

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- This formula says that the error in x_{n+1} is nearly proportional to the square of the error in x_n .
- When the initial error is sufficiently small, this shows that the error in the succeeding iterates will decrease very rapidly.

When to stop iterations?

Noting that $f(\alpha) = 0$, by Mean Value Theorem

$$f(x_n) = f(x_n) - f(\alpha) = f'(\xi_n)(x_n - \alpha).$$

Thus, error
$$\varepsilon_n = \alpha - x_n = -\frac{f(x_n)}{f'(x_n)}$$
 provided that x_n is close to α that $f'(x_n) \approx f'(\xi_n)$. This implies $\alpha - x_n \approx x_{n+1} - x_n$.

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■ This is the standard error estimation formula for Newton's method and it is usually fairly accurate. However, this formula is not valid if $f'(\alpha) = 0$.

Multiple Root

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$$f(x) = (x - \alpha)^m g(x)$$

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A zero of multiplicity 1 is called a simple root or a simple zero.

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• If α is a zero of f of multiplicity m. Then, one can write

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This implies

$$\mu(x) = \frac{(x - \alpha)^m g(x)}{m(x - \alpha)^{m-1} g(x) + (x - \alpha)^m g'(x)}$$
$$= (x - \alpha) \frac{g(x)}{mg(x) + (x - \alpha)g'(x)}$$

also has a zero at α . However, $g(\alpha) \neq 0$.

Newton's method for multiple roots

- Thus, $\mu'(\alpha) = \frac{1}{m} \neq 0$, hence α is called a simple zero of μ .
- Newton's method can be applied to $\mu(x)$ to give

$$x_{n+1} = x_n - \frac{f(x_n)f'(x_n)}{[f'(x_n)]^2 - f(x_n)f''(x_n)}$$

- This is called Newton's modified method. This has quadratic convergence regardless of multiplicity of the zeros of f.
- For a simple zero, the original Newton's method requires significantly low computations.

Newton versus Secant

Note that only one function evaluation is needed per step for the Secant method after x_2 has been determined. In contrast, each step of Newton's method requires an evaluation of both the function and its derivative.

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- Note that only one function evaluation is needed per step for the Secant method after x_2 has been determined. In contrast, each step of Newton's method requires an evaluation of both the function and its derivative.
- Newton's method or the Secant method is often used to refine an answer obtained by another technique, such as the bisection method, since these methods requires good first approximation but generally give rapid convergence.