

Chapter 8

Nonlinear algebraic equations

8.1 Nonlinear scalar equation

Fixed point iteration

We seek the solution $x \in I = [a, b] \subset \mathbb{R}$ of the equation

$$f(x) = 0, \quad (8.1)$$

with $f : I \rightarrow \mathbb{R}$ a nonlinear function, for which we can formulate the following fixed point iteration:

$$x^{(k+1)} = g(x^{(k)}) = x^{(k)} + \alpha f(x^{(k)}) \quad (8.2)$$

The fixed point iteration (8.2) converges to a unique solution $x = g(x)$, corresponding to $f(x) = 0$, if the function $g : I \rightarrow I$ is a *contraction mapping*, meaning that there exists a constant $L_g < 1$, such that

$$|g(x) - g(y)| \leq L_g |x - y|, \quad (8.3)$$

for all $x, y \in I$, where L_g is the *Lipschitz constant* of $g(x)$, a *Lipschitz continuous* function for $x \in I$.

Convergence of the fixed point iteration (8.2) is proven similar to the case of a linear system of equations (7.12). For any $k > 1$, we have that

$$|x^{(k+1)} - x^{(k)}| = |g(x^{(k)}) - g(x^{(k-1)})| \leq L_g |x^{(k)} - x^{(k-1)}| \leq L_g^k |x^{(1)} - x^{(0)}|,$$

and for $m > n$,

$$\begin{aligned} |x^{(m)} - x^{(n)}| &= |x^{(m)} - x^{(m-1)}| + \dots + |x^{(n+1)} - x^{(n)}| \\ &\leq (L_g^{m-1} + \dots + L_g^n) |x^{(1)} - x^{(0)}|, \end{aligned}$$

so that for $L_g < 1$, we have that

$$\lim_{n \rightarrow \infty} |x^{(n)} - x^{(n-1)}| = 0, \quad (8.4)$$

which implies that there exists an $x \in \mathbb{R}$ such that

$$x = \lim_{n \rightarrow \infty} x^{(n)}, \quad (8.5)$$

since $\{x^{(n)}\}_{n=1}^{\infty}$ is a Cauchy sequence in the complete space \mathbb{R} .

Uniqueness follows from assuming that there exists another solution $y \in \mathbb{R}$ such that $y = g(y)$, which leads to a contradiction, since

$$|x - y| = |g(x) - g(y)| \leq L_g |x - y| < |x - y|. \quad (8.6)$$

Thus x is the unique solution to the equation $x = g(x)$.

Rate of convergence

We are not only interested in *if* an iterative method converges, but also *how fast*, that is the *rate of convergence*. We say that a sequence of approximate solutions $\{x^{(k)}\}_{k=1}^{\infty}$ converges with *order* p to the exact solution x , if

$$\lim_{k \rightarrow \infty} \frac{|x - x^{k+1}|}{|x - x^{k+1}|^p} = C, \quad C > 0, \quad (8.7)$$

where $p = 1$ corresponds to a linear order of convergence, and $p = 2$ a quadratic order of convergence. We can approximate the rate of convergence by extrapolation, so that

$$p \approx \frac{\log \frac{|x^{k+1} - x^k|}{|x^k - x^{k-1}|}}{\log \frac{|x^k - x^{k-1}|}{|x^{k-1} - x^{k-2}|}} \quad (8.8)$$

which is useful in practice when the exact solution is not available.

Newton's method

The analysis above suggests that (8.2) converges linearly with the Lipschitz constant L_g , since

$$|x - x^{(k+1)}| = |g(x) - g(x^{(k)})| < L_g |x - x^{(k)}| \quad (8.9)$$

Algorithm 11: Newton's method

Given initial approximation $x^{(0)}$
while $|f(x^{(k)})| \geq TOL$ **do**
 $x^{(k+1)} = x^{(k)} - f'(x^{(k)})^{-1} f(x^{(k)})$
end

so that the error $e^{(k)} = x - x^{(k)}$ decreases linearly as $|e^{(k+1)}| < L_g |e^{(k)}|$.

Although, by the choice $\alpha = -f'(x^{(k)})^{-1}$, the fixed point iteration (8.2) exhibits quadratic convergence for $x^{(k)}$ close to x , the exact solution to the equation $f(x) = 0$. This is Newton's method.

The quadratic convergence of Newton's method follows from *Taylor's formula*, with $f(x)$ expanded around $x^{(k)}$, with $\xi \in I$,

$$0 = f(x) = f(x^{(k)}) + f'(x^{(k)})(x - x^{(k)}) + \frac{1}{2}f''(\xi)(x - x^{(k)})^2. \quad (8.10)$$

We divide by $f'(x^{(k)})$ to get

$$x - (x^{(k)} - f'(x^{(k)})^{-1} f(x^{(k)})) = -\frac{1}{2}f'(x^{(k)})^{-1} f''(\xi)(x - x^{(k)})^2, \quad (8.11)$$

so that for the error $e^{(k)} = x - x^{(k)}$,

$$|e^{(k+1)}| = \frac{1}{2}|f'(x^{(k)})^{-1} f''(\xi)| |e^{(k)}|^2, \quad (8.12)$$

which displays the quadratic convergence for $x^{(k)}$ close to x .

8.2 Systems of nonlinear equations

Fixed point iteration for nonlinear systems

Now consider systems of nonlinear equations: find $x = (x_1, \dots, x_n)^T \in \mathbb{R}^n$, such that

$$f(x) = 0, \quad (8.13)$$

with $f = (f_1, \dots, f_n)^T : \mathbb{R}^n \rightarrow \mathbb{R}^n$. We can formulate a fixed point iteration $x^{(k+1)} = g(x^{(k)})$, with $g : \mathbb{R}^n \rightarrow \mathbb{R}^n$, just as in the case of the scalar problem.

Existence of a unique solution to the algorithm follows by the Banach fixed point theorem.

Theorem 13 (Banach fixed point theorem). *The fixed point iteration of Algorithm 12 converges to a unique solution if $\|1 + \alpha L_f\| < 1$, with L_f the Lipschitz constant of the function $f(x)$.*

Algorithm 12: Newton's method for systems of nonlinear equations

Given initial approximation $x^{(0)}$
while $\|f(x^{(k)})\| \geq TOL$ **do**
 | $x^{(k+1)} = x^{(k)} + \alpha f(x^{(k)})$
end

Proof. For $k > 1$ we have that

$$\begin{aligned} \|x^{(k+1)} - x^{(k)}\| &= \|x^{(k)} - x^{(k-1)} + \alpha(f(x^{(k)}) - f(x^{(k-1)}))\| \\ &\leq (1 + \alpha L_f) \|x^{(k)} - x^{(k-1)}\| \leq (1 + \alpha L_f)^k \|x^{(1)} - x^{(0)}\|, \end{aligned}$$

and for $m > n$,

$$\begin{aligned} \|x^{(m)} - x^{(n)}\| &= \|x^{(m)} - x^{(m-1)}\| + \dots + \|x^{(n+1)} - x^{(n)}\| \\ &\leq ((1 + \alpha L_f)^{m-1} + \dots + (1 + \alpha L_f)^n) \|x^{(1)} - x^{(0)}\|. \end{aligned}$$

For $(1 + \alpha L_f) < 1$, $\{x^{(n)}\}_{n=1}^{\infty}$ is a Cauchy sequence since $\|x^{(m)} - x^{(n)}\| \rightarrow 0$, for $n \rightarrow \infty$, which implies that there exists an $x \in \mathbb{R}^n$ such that

$$x = \lim_{n \rightarrow \infty} x^{(n)}, \quad (8.14)$$

since \mathbb{R}^n is a complete vector space. Uniqueness follows from assuming that there exists another solution $y \in \mathbb{R}^n$ such that $f(y) = 0$, which leads to a contradiction, since

$$\|x - y\| = \|x - y + \alpha(f(x) - f(y))\| \leq (1 + \alpha L_f) \|x - y\| < \|x - y\|. \quad (8.15)$$

Thus x is the unique solution to the equation $f(x) = 0$. \square

Newton's method for nonlinear systems

The *Jacobian matrix* $f'(x)$ is defined as

$$f'(x) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}, \quad (8.16)$$

which we use to formulate Newton's method for systems of equations.

The quadratic convergence of Newton's method follows from *Taylor's formula* in \mathbb{R}^n , with $f(x)$ expanded around $x^{(k)}$,

$$0 = f(x) = f(x^{(k)}) + f'(x^{(k)})(x - x^{(k)}) + \frac{1}{2}(x - x^{(k)})^T f''(\xi)(x - x^{(k)}),$$

Algorithm 13: Newton's method for systems of nonlinear equations

Given initial approximation $x^{(0)}$
while $\|f(x^{(k)})\| \geq TOL$ **do**
 | Solve $f'(x^{(k)})\Delta x^{(k+1)} = -f(x^{(k)})$ ▷ solve for $\Delta x^{(k+1)}$
 | $x^{(k+1)} = x^{(k)} + \Delta x^{(k+1)}$ ▷ update by $\Delta x^{(k+1)}$
end

where $f''(x^{(k)})$ is the *Hessian*. We have that

$$x - (x^{(k)} - f'(x^{(k)})^{-1}f(x^{(k)})) = -\frac{1}{2}f'(x^{(k)})^{-1}(x - x^{(k)})^T f''(\xi)(x - x^{(k)}),$$

with $f'(x^{(k)})^{-1}$ the inverse of the Jacobian, so that,

$$\|e^{(k+1)}\| = \frac{1}{2}\|f'(x^{(k)})^{-1}f''(\xi)\| \|e^{(k)}\|^2, \quad (8.17)$$

for the error $e^{(k)} = x - x^{(k)}$, which shows quadratic convergence for $x^{(k)}$ close to x .