Notes on Nonlinear Newton Iterations at a Typical Singularity

Given a nonlinear n-vector-valued function f(x) of an n-vector x, a solution z of the equation f(z) = o is often sought by means of *Newton's Iteration* $x_{k+1} := Nf(x_k)$ wherein *Newton's*

Iterating Function $Nf(x) := x - f'(x)^{-1} \cdot f(x)$. Here $f'(x) := \partial f(x)/\partial x$ is the n-by-n Jacobian Matrix of first partial derivatives; it is usually presumed adequately differentiable and invertible at all x in some open neighborhood of the desired solution z, and then

 $Nf(x) - z = f'(x)^{-1} \cdot (f''(x) \cdot (x-z) + \dots) \cdot (x-x) \approx O(x-z)^2 \text{ as } x \to z.$

Those presumptions imply at least *Quadratic Convergence*. This happens almost always.

The iteration's behavior is not so easy to characterize in some singular situations. Among these the most common has $\det(f'(x)) = 0$ but $\partial \det(f'(x))/\partial x \neq 0$ at x = z. Thus $\det(f'(x)) = 0$ along some curve (n = 2) or (hyper)surface $(n \ge 3)$ that passes through z but does not intersect itself there, not even tangentially. Call this locus "\$". It divides some open neighborhood of z into two open regions in which $\det(f'(x))$ takes opposite signs. Nf(x) is either ambiguous or infinite on \$ because f'(x) is not invertible there. These notes explore the behavior of Newton's iteration when it converges to z but no iterate falls into \$. Convergence turns out to be *Linear* rather than quadratic, and iterates approaching z usually tend to avoid \$, as we shall see. Our conclusions are summarized on p. 3 and illustrated by examples on p. 4.

The Derivative row $h(x) := \partial det(f'(x))/\partial x$.

Jacobi's Formula for the derivative of a determinant says $d \det(B) = Trace(Adj(B) \cdot dB)$ wherein the Trace is the sum of all diagonal elements, and Adj(B) is the *Classical Adjoint* or *Adjugate*: Adj(B) := det(B) $\cdot B^{-1}$ when det(B) $\neq 0$ and is otherwise defined by the continuity of what turns out to be a polynomial function of the elements of B defined by $B \cdot Adj(B) \equiv Adj(B) \cdot B \equiv det(B) \cdot I$ in general. There are other equivalent definitions of Adj(B) in terms of determinants or the *Characteristic Polynomial* of B, but all we need from them is this fact: $Adj(B) \neq O$ just when the n-by-n matrix B has $Rank(B) \ge n-1$, and then $Rank(Adj(B)) = n - (n-1) \cdot (n - Rank(B))$. Jacobi's formula is derived at <<wr/>www.cs.berkeley.edu/~wkahan/MathH110/jacobi.pdf>.

Jacobi's formula says $d \det(f'(x)) = \operatorname{Trace}(\operatorname{Adj}(f'(x)) \cdot f''(x) \cdot dx)$ wherein the second derivative f'' is a bilinear operator that maps n-vectors y and z each linearly to an n-vector $f''(x) \cdot y \cdot z$ that, in general, varies nonlinearly with x and, if continuously, $f''(x) \cdot y \cdot z = f''(x) \cdot z \cdot y$. In particular $f''(x) \cdot dx$ is a matrix whose every element depends linearly upon column dx, so there must be some row n-vector $h'(x) = \partial \det(f'(x))/\partial x$ satisfying $d \det(f'(x)) = h'(x) \cdot dx$ for all dx; and if $h'(z) \neq o'$ then the equation of the plane tangent to \$ at z is $h'(z) \cdot (x-z) = 0$.

We assume henceforth that $h(z) \neq o$ and det(f'(z)) = 0, whence Rank(f'(z)) = n-1 and hence Rank(Adj(f'(z))) = 1, so $Adj(f'(z)) = w \cdot v$ for two nonzero vectors satisfying $f'(z) \cdot w = o$ and $v \cdot f'(z) = o$. Consequently

 $h^{(z)} \cdot dx = \operatorname{Trace}(\operatorname{Adj}(f'(z)) \cdot f''(z) \cdot dx) = \operatorname{Trace}(w \cdot v^{\cdot} f''(z) \cdot dx) = \operatorname{Trace}(v^{\cdot} f''(z) \cdot w \cdot dx)$ and therefore $h^{(z)} = v^{\cdot} \cdot f''(z) \cdot w = \partial \det(f'(x)) / \partial x$ at x = z. We shall need this formula later. A *Taylor Series* expansion of f will be presumed valid for x in some open neighborhood of z : $f(x) = o + f'(z) \cdot (x-z) + \frac{1}{2} f''(z) \cdot (x-z) \cdot (x-z) + \frac{1}{2} f'''(z) \cdot (x-z) \cdot (x-z) \cdot (x-z) + \dots$

$$f'(x) = f'(z) + f''(z) \cdot (x-z) + \frac{1}{2} f'''(z) \cdot (x-z) \cdot (x-z) + \dots; \quad \det(f'(z)) = 0; \quad \operatorname{Adj}(f'(z)) = w \cdot v^{*}.$$
$$\det(f'(x)) = h^{*} \cdot (x-z) + \dots \text{ wherein } h^{*} := v^{*} \cdot f''(z) \cdot w \neq o^{*}.$$

Needed next is a Taylor series expansion for Newton's iterating function:

$$Nf(x) = x - f'(x)^{-1} \cdot f(x)$$

= $z + \frac{1}{2} (f' + f'' \cdot (x-z) + \frac{1}{2} f''' \cdot (x-z) \cdot (x-z) + \dots)^{-1} \cdot (f'' \cdot (x-z) + \frac{2}{3} f''' \cdot (x-z) \cdot (x-z) + \dots) \cdot (x-z)$
in which all derivatives of $f(x)$ are evaluated at $x = z$. This is too messy. It must be simplified.

A Change of Coordinates.

By a process akin to *Gaussian Elimination* with *Pivotal Exchanges* of both rows and columns, we can obtain a diagonal matrix $L^{-1} \cdot f'(z) \cdot U^{-1} = M := \begin{bmatrix} I & o \\ o' & 0 \end{bmatrix}$ using suitably permuted lower- and upper-triangular matrices L and U. These figure in a simplifying change of variables from x to $u := U \cdot (x-z)$. Let $g(u) := L^{-1} \cdot f(z + U^{-1} \cdot u)$ so that Newton's iterating function for g is $Ng(u) := u - g'(u)^{-1} \cdot g(u) = U \cdot (Nf(z + U^{-1} \cdot u) - z)$. In other words, the iteration $x_{k+1} := Nf(x_k)$ starting from x_0 is mimicked by the iteration $u_{k+1} := Ng(u_k)$ starting from $u_0 := U \cdot (x_0 - z)$ in so far as every $u_k = U \cdot (x_k - z)$. Iterates $x_k \to z$ just as fast (if at all) as $u_k \to o$.

Thus no generality is lost by assuming z = o = f(o), $f'(o) = M := \begin{bmatrix} I & o \\ o' & 0 \end{bmatrix}$, $Adj(f'(o)) = \begin{bmatrix} O & o \\ o' & 1 \end{bmatrix} = v \cdot v^*$ in which $w^* = v^* = [o^* \quad 1]$, and the foregoing Taylor series for Nf(x) is simplified to ... $Nf(x) = \frac{1}{2}(M + f'' \cdot x + \frac{1}{2}f''' \cdot x \cdot x + \dots)^{-1} \cdot (f'' \cdot x + \frac{2}{3}f''' \cdot x \cdot x + \dots) \cdot x$ $= \frac{1}{2}x - \frac{1}{2}(M + f'' \cdot x + \frac{1}{2}f''' \cdot x \cdot x + \dots)^{-1} \cdot (M - \frac{1}{6}f''' \cdot x \cdot x + \dots) \cdot x$ in which all derivatives of f(x) are evaluated at x = 0. Further are evaluated of x = 0.

in which all derivatives of f(x) are evaluated at x = o. Further progress requires a partition of $f'' \cdot x = \begin{bmatrix} Xx & C \cdot x \\ x' \cdot R' & h' \cdot x \end{bmatrix}$ in which Xx is an (n–1)-by-(n–1) matrix whose every element is a linear function of x. Similarly for the column C·x, the row x`·R` and the scalar h`·x. In fact h`·x = v`·f''·x·v = v`·f''·v·x = [v`·R` h`·v]·x so h` = [v`·R` h`·v] = $\partial \det(f'(x))/\partial x$ at x = o as

 $h \cdot x = v \cdot f'' \cdot x \cdot v = v \cdot f'' \cdot v \cdot x = [v \cdot R' \quad h \cdot v] \cdot x$ so $h' = [v \cdot R' \quad h' \cdot v] = \partial \det(f'(x))/\partial x$ at x = o as expected, and $h' \neq o'$ has been assumed. Later much of our analysis will be affected by whether the last element of h', namely h' \cdot v, is zero.

Let
$$\begin{bmatrix} \mathbf{W} & \boldsymbol{\mu} \cdot \mathbf{c} \\ \boldsymbol{\mu} \cdot \mathbf{r'} & \boldsymbol{\mu} \end{bmatrix} := f'(\mathbf{x})^{-1} = \left(\mathbf{M} + f'' \cdot \mathbf{x} + \frac{1}{2}f''' \cdot \mathbf{x} \cdot \mathbf{x} + \dots\right)^{-1} = \left(\begin{bmatrix} \mathbf{I} + \mathbf{X}\mathbf{x} & \mathbf{C} \cdot \mathbf{x} \\ \mathbf{x'} \cdot \mathbf{R'} & \mathbf{h'} \cdot \mathbf{x} \end{bmatrix} + \frac{1}{2}f''' \cdot \mathbf{x} \cdot \mathbf{x} + \dots\right)^{-1}.$$

Here a process akin to Gaussian elimination provides estimates

$$\begin{split} \mu(\mathbf{x}) &= \det(\mathbf{I} + \mathbf{X}\mathbf{x} + O(\mathbf{x})^2) / \det(f'(\mathbf{x})) = 1 / (\mathbf{h} \cdot \mathbf{x} + O(\mathbf{x})^2), \\ \operatorname{column} \ \mathbf{c}(\mathbf{x}) &= -(\mathbf{I} - \mathbf{X}\mathbf{x}) \cdot \mathbf{C} \cdot \mathbf{x} + O(\mathbf{x})^2, \\ \operatorname{row} \ \mathbf{r}^*(\mathbf{x}) &= -\mathbf{x}^* \cdot \mathbf{R}^* \cdot (\mathbf{I} - \mathbf{X}\mathbf{x}) + O(\mathbf{x})^2, \\ \operatorname{matrix} \ \mathbf{W}(\mathbf{x}) &= \mathbf{I} - \mathbf{X}\mathbf{x} + \mu(\mathbf{x}) \cdot \mathbf{c}(\mathbf{x}) \cdot \mathbf{r}^*(\mathbf{x}) + O(\mathbf{x})^2 \quad \text{as} \ \mathbf{x} \to \mathbf{o}. \end{split}$$

These estimates produce an estimate of $Nf(x) = \frac{1}{2} \begin{bmatrix} I - W & o \\ -U + r' & 1 \end{bmatrix} \cdot x + \mu(x) \cdot O(x)^3$.

What happens next depends upon whether x approaches z (= 0) too nearly tangentially to the surface \$ on which det(f'(x)) = 0. If so, $\mathbf{h} \cdot \mathbf{x} = O(x)^2$, and then $\mu(x) = 1/O(x)^2$ can be arbitrarily big, so big that the computation of $f'(x)^{-1} \cdot f(x)$ and thus Nf(x) malfunctions because of roundoff if not division by zero.

To avoid that malfunction we must keep $|\mathbf{h} \cdot \mathbf{x}| >> O(\mathbf{x})^2$, though $\mathbf{h} \cdot \mathbf{x} = O(\mathbf{x})$, so that terms like $\mu(x) \cdot O(x)^2$ stay no bigger than O(x) as $x \to 0$. Under these circumstances W(x) = I + O(x)and then $Nf(x) = \frac{1}{2}([-\mu \cdot \mathbf{\hat{r}} \quad 1] \cdot x) \cdot v + O(x)^2 = O(x) \cdot v + O(x)^2$. Thus, if Nf(x) is not $O(x)^2$ it is likely to take the form $Nf(x) = \beta \cdot v + O(\beta)^2$ for some tiny scalar $\beta = O(x)$. Whether these circumstances persist and avoid malfunctions depends upon whether the last element h v of h is zero.

If, as is most likely, $\mathbf{h} \cdot \mathbf{v} \neq 0$ then iterates of the form $\mathbf{x} = \mathbf{\beta} \cdot \mathbf{v} + O(\mathbf{\beta})^2$ for sufficiently tiny nonzero scalars β turn into $Nf(x) = \frac{1}{2}\beta \cdot v + O(\beta)^2 \approx \frac{1}{2}x$ because $\mu(x) = 1/(\beta \cdot h \cdot v + O(\beta)^2)$ is not too tiny, and then convergence is linear with rate log(2) and iterates approach z along a direction v that is *Transverse* (not tangential) to the surface \$.

In the unlikely case that h v = 0 the iteration's behavior is difficult to predict because, although iterates x tend often to come close to the form $x = \beta \cdot v + O(\beta)^2$, it is too nearly tangential to \$.

Summary.

If f(z) = 0 and det(f'(z)) = 0 and Newton's iteration is started close enough to z but not too much closer to the locus \$ on which det(f'(x)) = 0, the iteration's convergence to z is linear with rate log(2) provided a technical condition $(\partial \det(f'(x))/\partial x \text{ at } x = z) \cdot \operatorname{Adj}(f'(z)) \neq 0$ is satisfied by the second derivative f''(z), as is usually the case. This technical condition can be described in terms of nonzero null-vectors $v \neq o = v \cdot f'(z)$ and $w \neq o = f'(z) \cdot w$ of singular matrix f'(z): it is that $v \cdot f''(z) \cdot w \cdot w \neq 0$. And then iterates $x_{k+1} := x_k - f'(x_k)^{-1} \cdot f(x_k)$ approach the desired zero z very nearly like $z + (\beta/2^k) \cdot w$ for some small nonzero scalar constant β .

Example p :

For column 2-vector arguments x let column 2-vector $p(x) := a + B \cdot x + \left| \frac{x' \cdot C_1 \cdot x}{x' \cdot C_2 \cdot x} \right| / 2$ in which

 $a = \begin{bmatrix} -15/2 \\ 11 \end{bmatrix}, B = \begin{bmatrix} 1 & -2 \\ 0 & 2 \end{bmatrix}, C_1 = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}, C_2 = \begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix}$. Now $p(\begin{bmatrix} 17/2 \\ -9/2 \end{bmatrix}) = o$ and $p'(\begin{bmatrix} 17/2 \\ -9/2 \end{bmatrix}) = \begin{bmatrix} 1 & 3 \\ 7 & 19 \end{bmatrix}/2$ is nonsingular, so Newton's iteration converges quadratically to this zero of p as expected. But $\mathbf{p}\begin{pmatrix} \begin{bmatrix} 7\\-4 \end{bmatrix}) = \mathbf{o} \text{ has a singular } \mathbf{p'}\begin{pmatrix} \begin{bmatrix} 7\\-4 \end{bmatrix}) = \begin{bmatrix} 0 & 0\\2 & 7 \end{bmatrix}; \text{ now } \mathbf{v} = \begin{bmatrix} 1 & 0 \end{bmatrix}, \mathbf{w} = \begin{bmatrix} 7\\-2 \end{bmatrix} \text{ and } \mathbf{v} \cdot f''(\mathbf{z}) \cdot \mathbf{w} \cdot \mathbf{w} = 5 \neq 0,$ so Newton's iterates x_k tend to this "double" zero z of p linearly like $z + (\beta/2^k) \cdot w$. Try it!

Newton's iterating function $Np(x) := x - p'(x)^{-1} \cdot p(x)$ may tend to infinity as x tends to the parabola on which det(p'(x)) = 0. The equation $-det(p'(\begin{bmatrix} \xi \\ n \end{bmatrix})) = (\xi + \eta - 7)^2 + 5\xi - 51 = 0$ of the parabola is solved by its parametrization: $\xi = \xi(\tau) := 7 + 8\tau - 5\tau^2$ and $\eta = \eta(\tau) := 5\tau^2 - 3\tau - 4$. Np(x) becomes indeterminate at two points on this parabola: One is the "double" zero $z = \begin{bmatrix} 7 \\ 4 \end{bmatrix}$ and the second is the point $x = \begin{bmatrix} 8.4 \\ -4.4 \end{bmatrix}$. As x approaches a third point $\begin{bmatrix} 338/45 \\ -188/45 \end{bmatrix}$ on the parabola, the direction of Np(x)–x = –p'(x)⁻¹·p(x) approaches tangency with the parabola; elsewhere than near these three points, the direction of Np(x)-x throws Newton's iterate Np(x) violently across the parabola as x approaches it.



Example q :

For column 2-vector arguments x let column 2-vector $q(x) := a + B \cdot x + \begin{vmatrix} x' \cdot C_1 \cdot x \\ x' \cdot C_2 \cdot x \end{vmatrix} / 2$ in which

$$\mathbf{a} = \begin{bmatrix} -6 \\ -4 \end{bmatrix}, \ \mathbf{B} = \begin{bmatrix} -2 & 0 \\ 0 & 2 \end{bmatrix}, \ \mathbf{C}_1 = \begin{bmatrix} 1 & 2 \\ 2 & 3 \end{bmatrix}, \ \mathbf{C}_2 = \begin{bmatrix} 2 & 3 \\ 3 & 4 \end{bmatrix}. \text{ Now } \mathbf{q}(\begin{bmatrix} -6 \\ 4 \end{bmatrix}) = \mathbf{0} \text{ and } \mathbf{q'}(\begin{bmatrix} -6 \\ 4 \end{bmatrix}) = \mathbf{0} \text{ has rank } \mathbf{0} \text{ so the so the so the so that } \mathbf{0} \text{ and } \mathbf{q'}(\begin{bmatrix} -6 \\ 4 \end{bmatrix}) = \mathbf{0} \text{ has rank } \mathbf{0} \text{ so the so t$$

analysis displayed above cannot explain the linear convergence of Newton's iteration to this zero z. Worse, q(x) = 0 all along the line whose equation is $\begin{bmatrix} 1 \\ 1 \end{bmatrix} \cdot x = -2$ and thereon, except at this zero z, q'(x) is a nonzero scalar multiple of $\begin{bmatrix} 1 & 1 \end{bmatrix} \begin{bmatrix} 1 & 1 \end{bmatrix}$, so v and w are nonzero scalar multiples of [1 -1] whence $v \cdot f''(z) \cdot w \cdot w = 0$. Consequently the analysis displayed above does not explain why Newton's iterates x_k converge to no zero of q on that line other than the zero

$$z = \begin{bmatrix} -6 \\ 4 \end{bmatrix}$$
, and converges to it like $z + (\beta/2^k) \cdot w$. Try it!