

## ROB 501 Handout: Grizzle

### Newton Raphson Algorithm

Let  $h : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be continuously differentiable, and satisfy

$$\det \left( \frac{\partial h}{\partial x}(x) \right) \neq 0 \quad \forall x \in \mathbb{R}^n$$

**Problem:** For  $y \in \mathbb{R}^n$  fixed, find a solution of  $y = h(x)$ ; i.e, find  $x^* \in \mathbb{R}^n$  s.t.  $y = h(x^*)$ . We note that this is equivalent to  $h(x^*) - y = 0$ . In other words, we are looking for a root of the equation  $h(x) - y = 0$ ,

**Approach:** Find a convergent sequence  $x_k \rightarrow x^*$  such that

$$\lim_{k \rightarrow \infty} h(x_k) - y = h(x^*) - y = 0$$

that is,  $x^* = \lim_{k \rightarrow \infty} x_k$  is a root of  $h(x) - y = 0$

**Idea:** Write  $x_{k+1} = x_k + \Delta x_k$ . We want

$$h(x_{k+1}) - y = h(x_k + \Delta x_k) - y \approx 0.$$

What should  $\Delta x_k$  look like ?

Apply Taylor's Theorem, to get

$$\begin{aligned} h(x_k) + \frac{\partial h}{\partial x}(x_k) \Delta x_k - y &\approx 0 \\ \therefore \frac{\partial h}{\partial x}(x_k) \Delta x_k &\approx -(h(x_k) - y) \\ \Delta x_k &\approx - \left( \frac{\partial h}{\partial x}(x_k) \right)^{-1} (h(x_k) - y) \end{aligned}$$

Recalling that  $x_{k+1} = x_k + \Delta x_k$ , we arrive at Newton's Algorithm,

$$x_{k+1} = x_k - \left( \frac{\partial h}{\partial x}(x_k) \right)^{-1} (h(x_k) - y)$$

In practice, the change in  $x_k$  given by  $\Delta x_k = - \left( \frac{\partial h}{\partial x}(x_k) \right)^{-1} (h(x_k) - y)$  is often too large. Hence, one uses the so-called Damped Newton Algorithm

$$x_{k+1} = x_k - \epsilon \left( \frac{\partial h}{\partial x}(x_k) \right)^{-1} (h(x_k) - y)$$

where  $\epsilon > 0$  provides step size control!

**Remark:** Looking ahead to our discussion of contraction mappings, let's rewrite the algorithm as the iteration of a mapping  $x_{k+1} = P(x_k)$

$$P(x) := x - \epsilon \left( \frac{\partial h}{\partial x}(x) \right)^{-1} (h(x) - y)$$

A solution of  $h(x) - y$  is a fixed point of  $P(x)$ . Indeed,

$$\begin{aligned} x^* &= P(x^*) \\ \Downarrow \\ x^* &= x^* - \epsilon \left( \frac{\partial h}{\partial x}(x^*) \right)^{-1} (h(x^*) - y) \\ \Downarrow \\ 0 &= -\epsilon \left( \frac{\partial h}{\partial x}(x^*) \right)^{-1} (h(x^*) - y) \\ \Downarrow \\ 0 &= (h(x^*) - y). \end{aligned}$$

It can be shown that  $P$  is a local contraction on an open ball around a solution of  $h(x) - y = 0$ .

**Example** Find the solution to the coupled NL equations

$$0 = h(x) = \begin{pmatrix} x_1 + 2x_2 - x_1(x_1 + 4x_2) - x_2(4x_1 + 10x_2) + 3 \\ 3x_1 + 4x_2 - x_1(x_1 + 4x_2) - x_2(4x_1 + 10x_2) + 4 \\ \sin(x_3)^7 + \frac{\cos(x_1)}{2} \\ x_4^3 - 2x_2^2 \sin(x_1) \end{pmatrix}$$

**Initial Guess:**  $x_0 = \begin{bmatrix} 7 \\ 8 \\ 9 \\ 10 \end{bmatrix}$

We do 16 iterations of Newton's Algorithm (a nonlinear root finding algorithm) and we obtain:

$$x^* = \begin{pmatrix} -2.25957308738366677539068499960 \\ 1.75957308738366677539068499960 \\ 189.50954100613333978330549312824 \\ -1.68458069860197189523093013800 \end{pmatrix}$$

**And the error is:**

$$h(x^*) = \begin{bmatrix} 3.6734198 \times 10^{-39} \\ 2.9387359 \times 10^{-39} \\ 1.2765134 \times 10^{-38} \\ -2.5915832 \times 10^{-32} \end{bmatrix}$$