

Now, let's assume $f'(x_*) \neq 0$

(1) Then, since $f'(x)$ is assumed continuous in a neighborhood of x_* , $\exists \delta_1 > 0$ s.t.

$$|f'(x)| > \frac{1}{2} \alpha \quad \forall |x - x_*| < \delta_1$$

(2) Since $f''(x)$ is also continuous,

$\exists \delta_2 > 0, M > 0$, s.t.

$$|f''(x)| \leq M \quad \forall |x - x_*| < \delta_2$$

Let $\delta_3 = \min(\delta_1, \delta_2)$. Then

$$\forall x, y \quad \left| \frac{f''(x)}{2f'(y)} \right| \leq C = \frac{M}{\alpha} (> 0)$$

$$\forall x, y \quad \text{s.t. } |x - x_*| \leq \delta_3 \text{ \& } |y - x_*| \leq \delta_3$$

Here is an outline of the proof.

Now, let's choose x_0 .

How do we pick x_0 ?

① Pick $x_0 \in \mathbb{R}$ s.t. $|x_0 - x_*| \leq \delta_3$

Then $|x_* - x_0| \leq \frac{1}{C}$

$\Rightarrow |x_* - x_1|$

Now, let's assume $|f'(x_*)| = \alpha \neq 0$.

(1) Since f' is continuous, $\exists \delta_1 > 0$ s.t.

$$|f'(x)| \geq \frac{\alpha}{2} \quad \forall |x - x_*| \leq \delta_1$$

(2) Since $f''(x)$ is continuous,

$\exists \delta_2 > 0, M > 0$ s.t.

$$|f''(x)| \leq M \quad \forall |x - x_*| \leq \delta_2$$

Let $\delta_3 = \min(\delta_1, \delta_2)$, Then $\exists 0 < \bar{C} < \infty$

s.t. $C(y, z) = \left| \frac{f''(y)}{2f'(z)} \right| < \bar{C}$

$$\forall |y - x_*| \leq \delta_3 \text{ \& } |z - x_*| \leq \delta_3$$

For N.M. we have

$$C_k = \left| \frac{f''(\eta_k)}{2f'(x_k)} \right| = C(\eta_k, x_k) \text{ w.}$$

$$|\eta_k - x_*| \leq |x_k - x_*|$$

Outline of convergence proof

It all comes down, as it must, to the choice of x_0 .

$$|x_1 - x_*| = C(\eta_0, x_0) |x_0 - x_*|^2$$

$$\textcircled{1} \text{ Choose } |x_0 - x_*| \leq \delta_3$$

$$\Rightarrow C(\eta_0, x_0) \leq \bar{C}$$

$$\textcircled{2} \text{ Let } \delta = \min(\delta_3, \frac{1}{2} \bar{C}) > 0$$

$$\Rightarrow \bar{C} |x_0 - x_*|^2 \leq \frac{1}{2} |x_0 - x_*|$$

$$\Rightarrow |x_1 - x_*| = C_0 |x_0 - x_*|^2 \leq \frac{1}{2} |x_0 - x_*| \leq \frac{1}{2} \delta$$

At the next step we have

$$|x_2 - x_*| = C(\eta_1, x_1) |x_1 - x_*|^2$$

$$\text{w. } |\eta_1 - x_*| \leq |x_1 - x_*| \leq \frac{1}{2} \delta < \delta$$

$$\Rightarrow C_1 = C(\eta_1, x_1) \leq \bar{C}$$

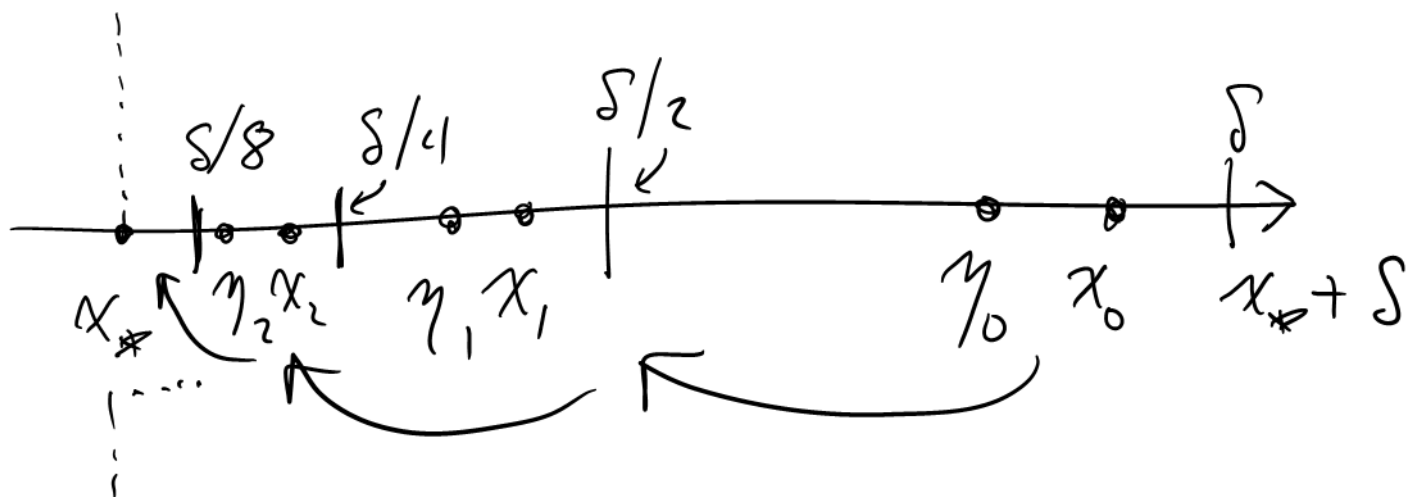
$$\Rightarrow |x_2 - x_*| \leq \frac{1}{2} |x_1 - x_*| \quad \text{by the same argument} \\ \leq \left(\frac{1}{2}\right)^2 \delta < \delta$$

and so on. At each iteration, η_k & x_k stay within the δ -neighborhood of x_* , & C_k can be bounded by \bar{C} .

Iterating this we have

$$|y_k - x_*| \leq |x_k - x_*| \leq \left(\frac{1}{2}\right)^k \delta$$

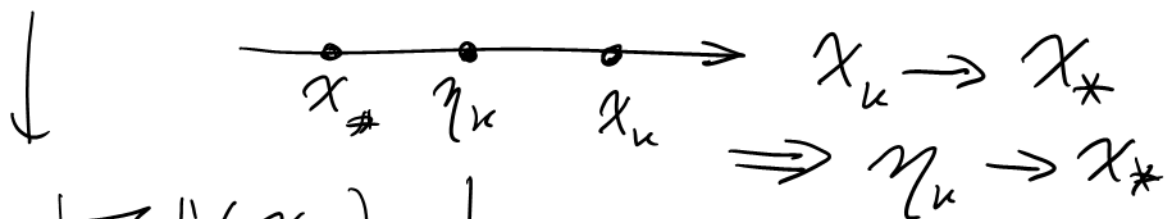
Hence, $y_k, x_k \rightarrow x_*$ as $k \rightarrow \infty$.



Defn A seq. $\{x_k\}$, w. $x_k \rightarrow x_*$, is said to converge to x_* w. at least order g if $\exists \{\varepsilon_k\}; \varepsilon_k \rightarrow 0$, and $\mu > 0$ s.t.
 $|x_k - x_*| \leq \varepsilon_k \quad k=0,1,\dots$ & $\lim_{k \rightarrow \infty} \frac{\varepsilon_{k+1}}{\varepsilon_k^g} \rightarrow \mu$

$$|x_{k+1} - x_*| = C_k |x_k - x_*|^2$$

$$\lim_{k \rightarrow \infty} \frac{|x_{k+1} - x_*|}{|x_k - x_*|^2} = \lim_{k \rightarrow \infty} \left| \frac{f''(\eta_k)}{2f'(x_k)} \right|$$



$$= \left| \frac{f''(x_*)}{2f'(x_*)} \right| = \mu \quad \text{i.e. } g=2.$$

Note If $f''(x_*) = 0$, then the defn above apparently breaks down.

This is silly since $\mu = 0$ reflects a super-quadratic rate of conv.

For N.M. when $f''(x_*) = 0$ & the

defn says "at least order g " (i.e. quadratic here.)

If $f''(x_*) = 0$, $g = 3$ (cubic)

Q: Will N.M. converge if $f'(x_*) = 0$?

Let's work a simple example.

Let $f(x) = x^p$; $p \geq 1$ & integer
 $f'(x) = p x^{p-1}$ [$x_* = 0$]

N.M. $x_{k+1} = x_k - \frac{x_k^p}{p x_k^{p-1}} = (1 - \frac{1}{p}) x_k$

i.e. $|x_{k+1} - x_*| = (1 - \frac{1}{p}) |x_k - x_*|$

This is only linear converge for $p > 1$
& for $p = 1$, converge to exact solution in 1 iteration for any x_0 .

Q: Why not go to a higher order approx. to f @ $x = x_k$?

N.M. $M(x; x_k) = f(x_k) + f'(x_k)(x - x_k)$

x_{k+1} satisfies $M(x_{k+1}; x_k) = 0$

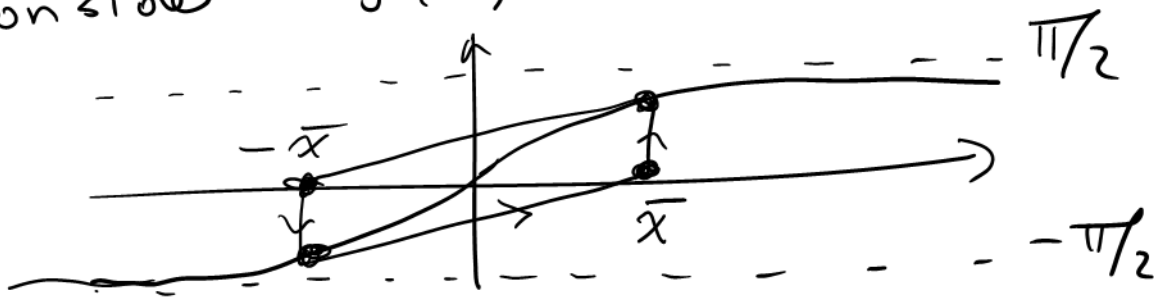
NNM: $Q(x; x_k) = f(x_k) + f'(x_k)(x - x_k) + \frac{1}{2} f''(x_k)(x - x_k)^2$

w. $Q(x_{k+1}; x_k) = 0$

- (i) Must solve quadratic eqn w. multiple roots; which is x_{n+1} ?
- (ii) Reg's higher deriv. info.
- (iii.) Do not generalize well to higher dimensions
- (iv) N.M. is pretty fast and almost certainly more robust.

A weakness of N.M. - x_0 must be suitably close for convergence.

Consider $f(x) = \arctan x$



$$f'(x) = \frac{1}{1+x^2} ; x_{n+1} = x_n - (1+x_n^2) \arctan x_n$$

Find \bar{x} s.t. $-\bar{x} = \bar{x} - (1+\bar{x}^2) \arctan \bar{x}$

$$h(\bar{x}) = 2\bar{x} - (1+\bar{x}^2) \arctan \bar{x}$$

$$h'(\bar{x}) = 2 - 2\bar{x} \arctan \bar{x} - 1$$

$$= 1 - 2\bar{x} \arctan \bar{x}$$

Use N.M $\bar{x} = 1.3917 \dots$

$$x_0 > \bar{x} \Rightarrow |x_n| \rightarrow \infty$$

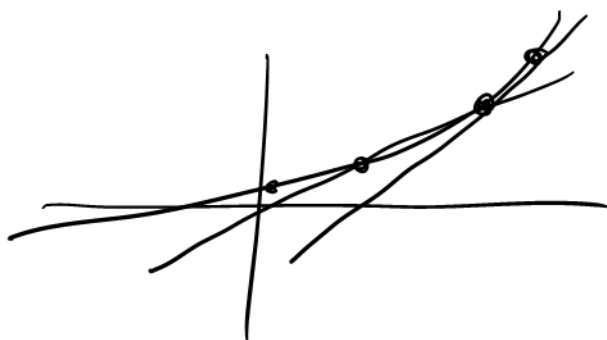
$$x_0 < \bar{x} \Rightarrow |x_n| \rightarrow 0$$

What if derivatives are hard to get?

Secant method

$$f'(x_n) \approx \frac{f(x_n) - f(x_{n-1})}{x_n - x_{n-1}}$$

$$x_{n+1} = x_n - f(x_n) \frac{x_n - x_{n-1}}{f(x_n) - f(x_{n-1})}$$



Need x_0 & x_1

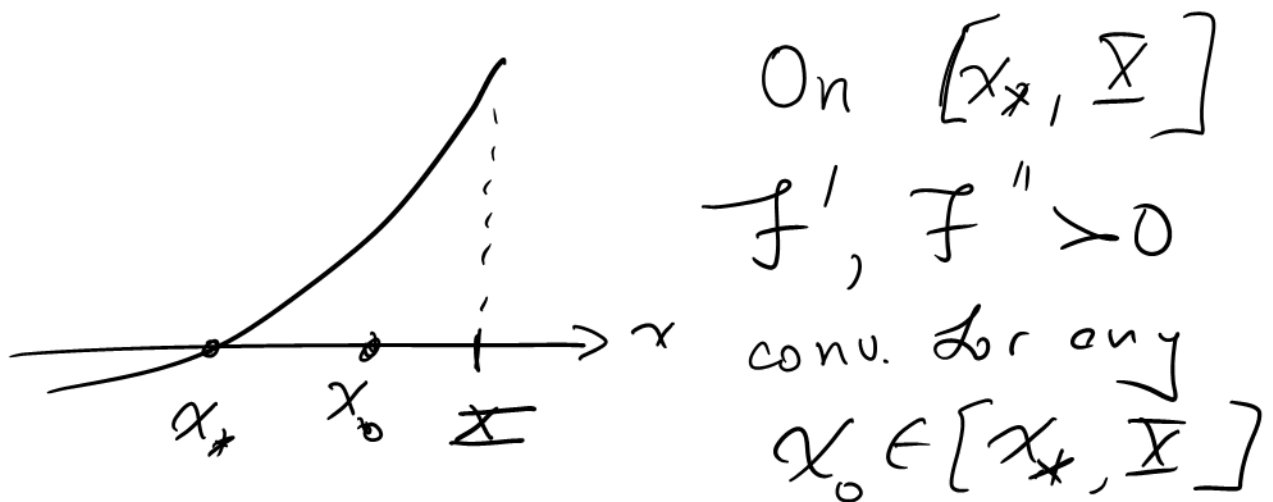
Then Use conditions very similar to N.M., S.M converges but slightly weaker. Instead of $g=2$,

$$g = \frac{1}{2} (1 + \sqrt{5}) \approx 1.6 \quad \left| \begin{array}{l} \text{super-linear} \end{array} \right.$$

S.M. generalizes in a very interesting way to systems of eqns through methods called quasi-Newton or Broyden schemes.

Read about direction As I implied, shows linear convergence w. $\mu = 1/2$.

In some cases, the "close enough" condition on x_0 can be relaxed



What if we have a system of eqns?

Find $\underline{x}_* = (x_{1*}, x_{2*}, \dots, x_{n*}) \in \mathbb{R}^n$ s.t.

$$\begin{aligned} f_1(x_{1*}, x_{2*}, \dots, x_{n*}) &= 0 & \underline{F} &= (f_1, \dots, f_n) \\ & \vdots & \underline{F}(\underline{x}_*) &= \underline{0} \\ f_n(x_{1*}, x_{2*}, \dots, x_{n*}) &= 0 & \underline{F}: \mathbb{R}^n &\rightarrow \mathbb{R}^n \end{aligned}$$

Approx. $\underline{F}(\underline{x})$ at $\underline{x} = \underline{x}_k$ by

$$\underline{M}(\underline{x}; \underline{x}_k) = \underline{F}(\underline{x}_k) + \underline{J}(\underline{x}_k)(\underline{x} - \underline{x}_k)$$

N.M. in \mathbb{R}^n : $\underline{M}(\underline{x}_{k+1}; \underline{x}_k) = \underline{0}$

$$\Leftrightarrow \underline{J}(\underline{x}_k)(\underline{x}_{k+1} - \underline{x}_k) = -\underline{F}(\underline{x}_k)$$

\underline{J} is a $\mathbb{R}^{n \times n}$ matrix $J_{ij} = \partial f_i / \partial x_j$