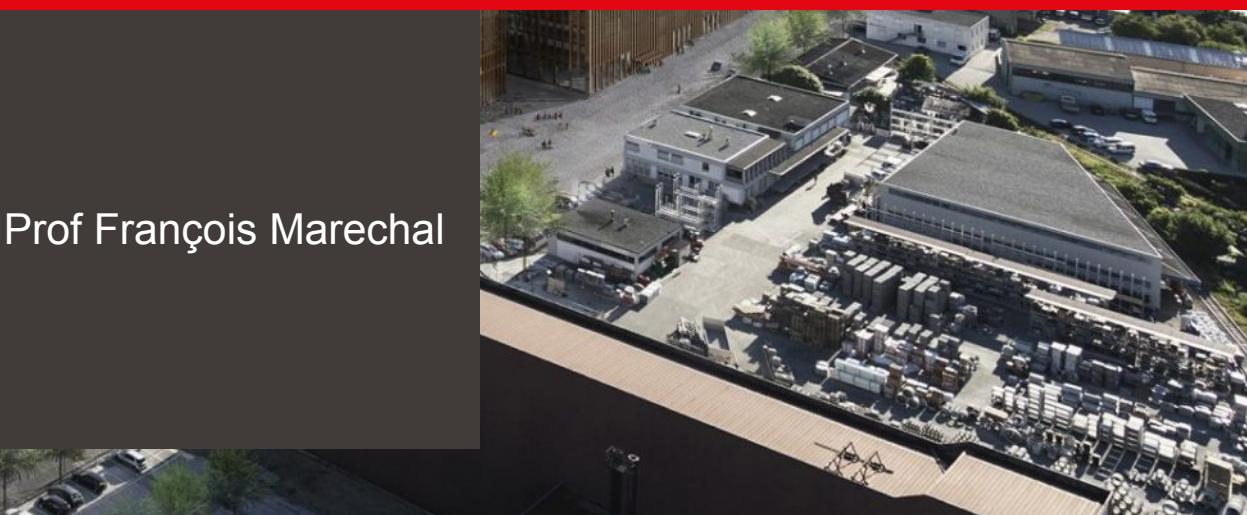


Solving Non linear equations



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- The goal : *Find the value of x^* such that $f(\pi_p, x^*) = 0$*
 - Calculation algorithm :
 - calculates the value of $f(\pi_p, x)$ for any value of x it receives
 - Solution algorithm that is searching for the value of x^*
 - Safe guard algorithm such that
 - x^* has a meaning : i.e. $x_{min} \leq x^* \leq x_{max}$

TAYLOR development to approximate any function

$$f(x) = f(x^0) + (x - x^0) \cdot f'(x^0) + \frac{(x - x^0)^2}{2!} \cdot f''(x^0) + \dots$$

First order limitation (approximate with a straight line) near x^0

$$f(x^0) + (x^* - x^0) \cdot f'(x^0) \approx f(x^*) = 0$$

$$x^1 = x^0 - \frac{f(x^0)}{f'(x^0)} \quad \rightarrow \quad x^{n+1} = x^n - \frac{f(x^n)}{f'(x^n)}$$



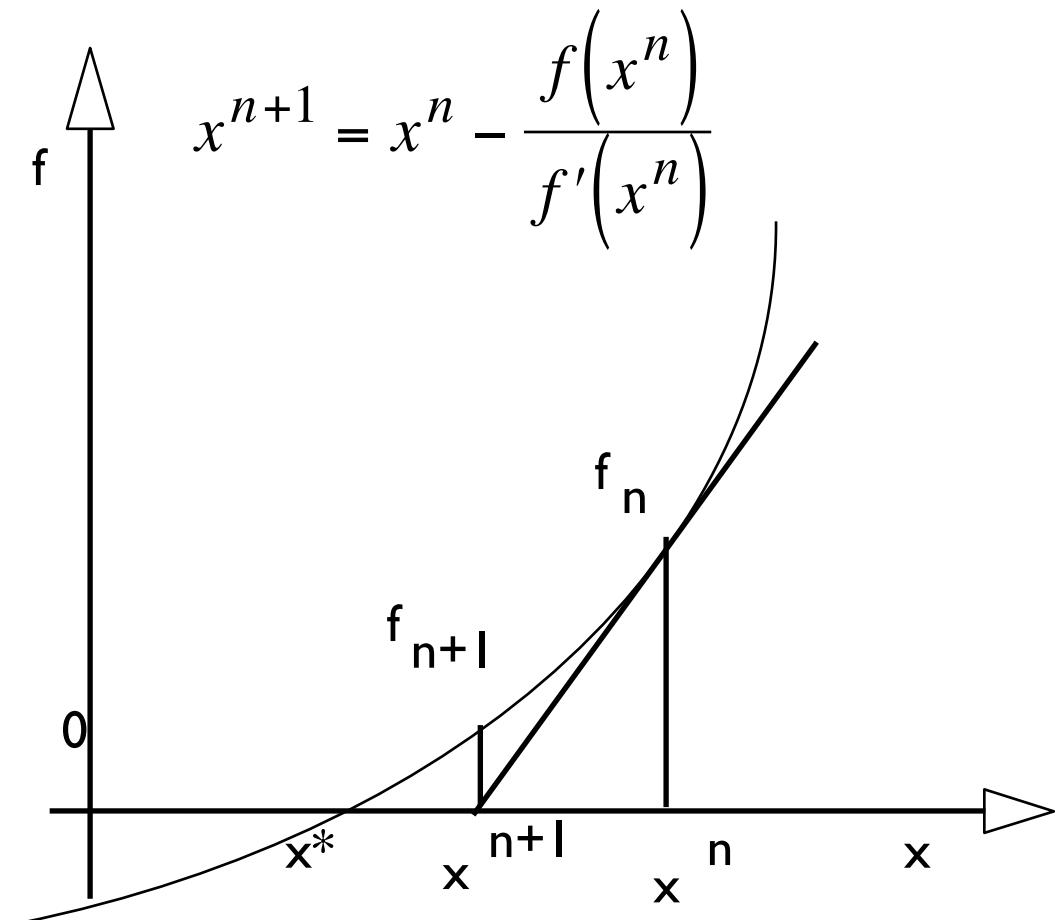
- Take x_0 = guess of x^* .
- Taylor development
 - $f(x) = f(x_0) + (x - x_0) f'(x_0) + 1/2 (x - x_0)^2 f''(x_0) + \dots$
- Limit to 1rst order :
 - $f(x_0) + (x^* - x_0) \cdot f'(x_0) \approx f(x^*) = 0$

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

- Repeat (Iterations) :
 - until $f(x)$ sufficiently close to 0



- tangent approximation
- intersection with $f=0$

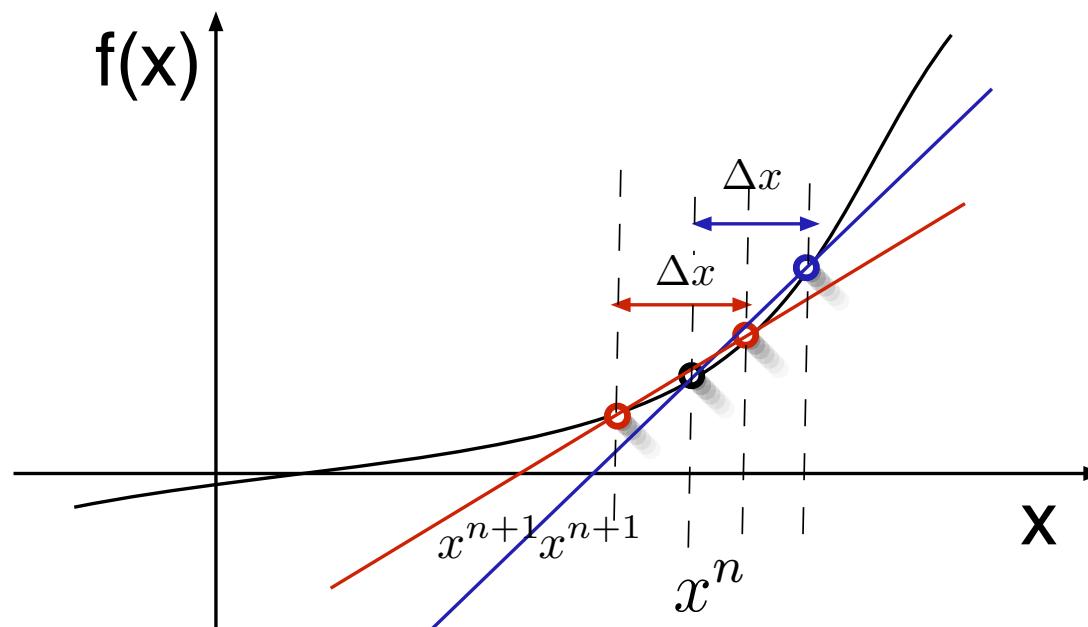


Forward calculation

$$f'(x^n) = \left[\frac{\delta f(x)}{\delta x} \right]_{x^n} = \frac{f(x^n) - f(x^n + \Delta x)}{\Delta x} \quad + 1 \text{ function evaluation}$$

Central value

$$f'(x^n) = \left[\frac{\delta f(x)}{\delta x} \right]_{x^n} = \frac{f(x^n - \frac{\Delta x}{2}) - f(x^n + \frac{\Delta x}{2})}{\Delta x} \quad + 2 \text{ function evaluations}$$



- Choose a small well chosen value

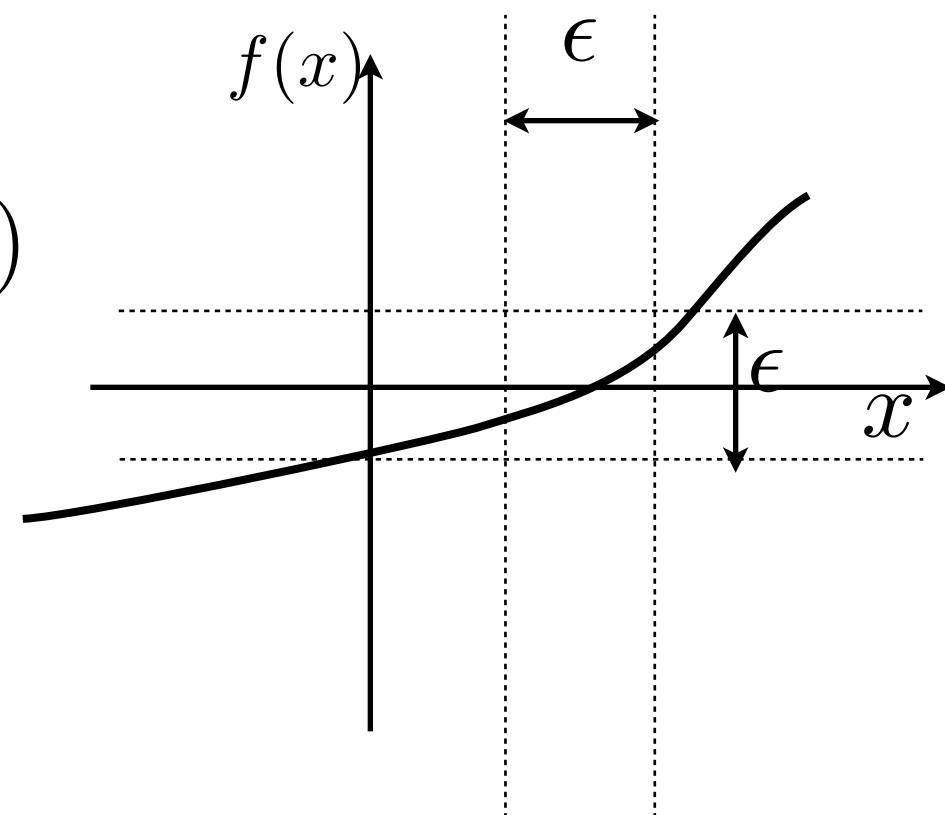
$$\epsilon = 1.0E - 06$$

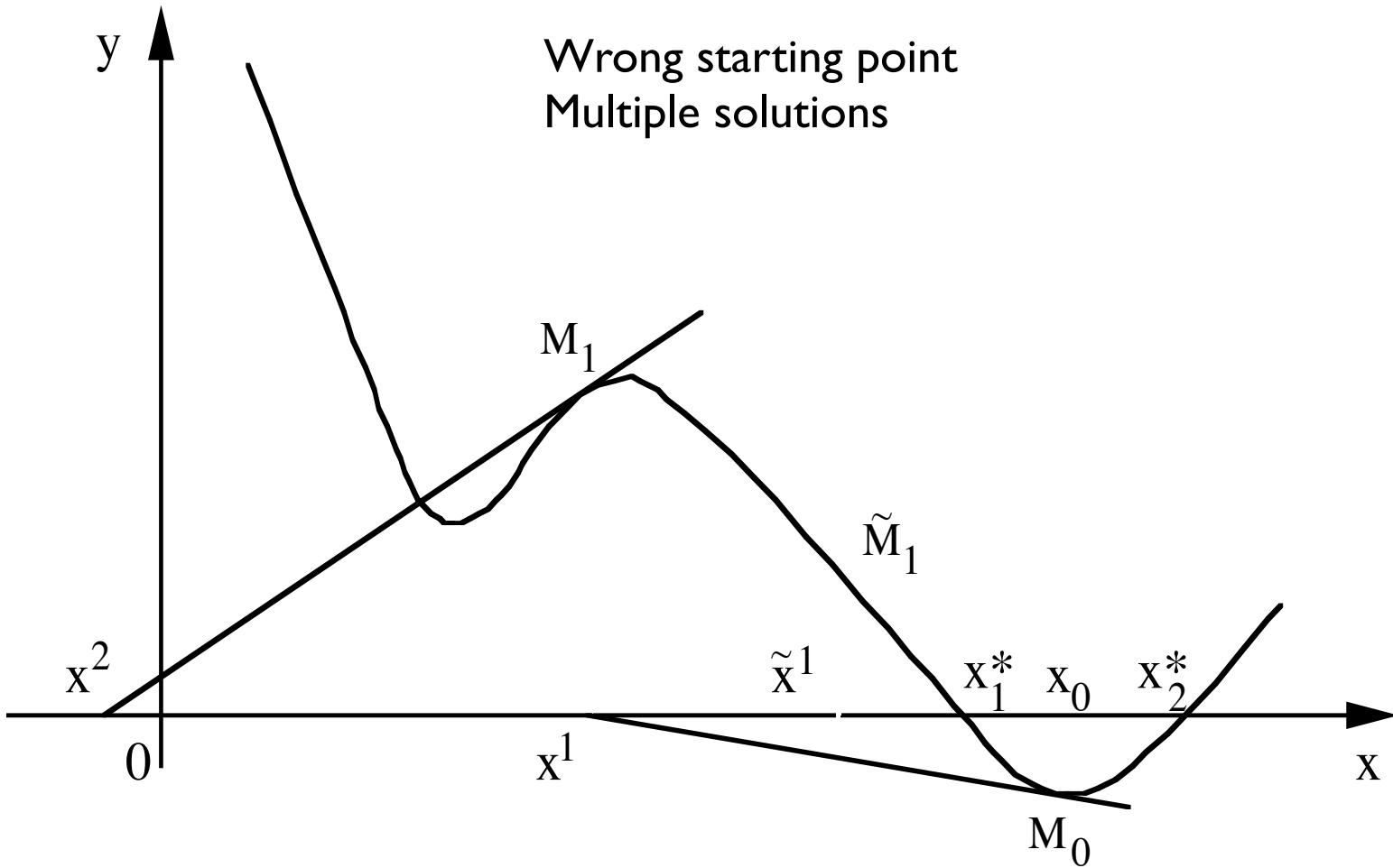
- Test variable variations

$$|x^{k+1} - x^k| \leq \epsilon * (1 + |x^k|)$$

- Test function variation

$$|f(x^k)| \leq \epsilon$$





EPFL Newton-Raphson : drawbacks

- Requires initial point
- Requires derivatives calculation
- divergence possible
- no step if tangent = 0



- Use relaxation factor $0 < q < 1$

$$x^{n+1} = x^n \cdot q + \tilde{x}^{n+1} \cdot (1 - q)$$

- $q=0 \Rightarrow$ full step

- Choice of q is critical ?

- convergence speed wrt robustness
 - $q \gg$ small steps and lots of iterations
 - $q \ll$ Newton step : direct convergence if linear problem

- Use past iterations to judge the need for relaxation :

$$\text{is } |f(x^{n+1})| < |f(x^n)| \text{ ?}$$



- 2nd order development
 - $f(x)=0=c+b(x-x_0)+a(x-x_0)^2$

$$x - x_0 = \frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

– a, b, c obtained from 3 values of f(x)

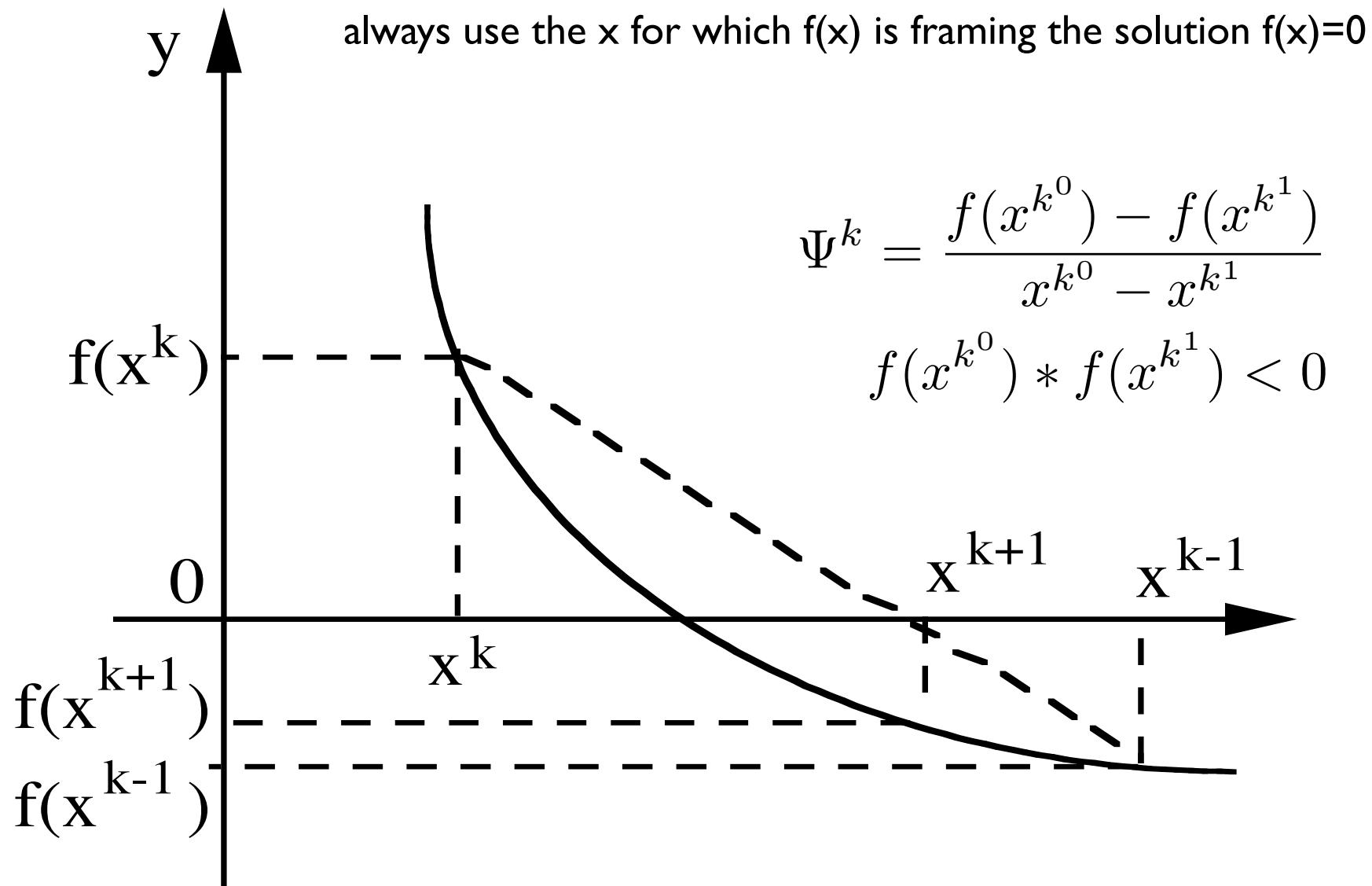
$$f(x) - f(x^k) = \frac{f(x^k) - f(x^{k-1})}{x^k - x^{k-1}} (x - x^k)$$

$$x^{k+1} = x^k - \psi^{-1} \cdot f(x^k)$$

$$\psi = \frac{f(x^k) - f(x^{k-1})}{x^k - x^{k-1}}$$

approximation of the derivative by the iteration step





- **Newton-Raphson n dimensions**

- Solve $F(X) = 0$ (Array)

$$f_1(x) = f_1(x_1, x_2, \dots, x_n) = 0$$

$$f_2(x) = f_2(x_1, x_2, \dots, x_n) = 0$$

...

$$f_n(x) = f_n(x_1, x_2, \dots, x_n) = 0$$

- Initial guess is given
 - $X_0 = \text{init}[x_1 \dots x_n]$



- Taylor development

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$$f_i(\underline{x}) = f_i(\underline{x}_0) + \sum_{j=1}^n \left(\frac{\partial f_i}{\partial x_j} \right) \Delta x_j = 0 \quad i = 1, n$$

Jacobian matrix
matrix of derivatives

function evaluated for \underline{x}^0

$$\begin{vmatrix}
 \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_j} & \dots & \frac{\partial f_1}{\partial x_n} \\
 \dots & & \dots & & \dots \\
 \frac{\partial f_i}{\partial x_1} & \dots & \frac{\partial f_i}{\partial x_j} & \dots & \frac{\partial f_i}{\partial x_n} \\
 \dots & & \dots & & \dots \\
 \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_j} & \dots & \frac{\partial f_n}{\partial x_n}
 \end{vmatrix} \cdot \begin{vmatrix} \Delta x_1 \\ \dots \\ \Delta x_j \\ \dots \\ \Delta x_n \end{vmatrix} + \begin{vmatrix} f_1(\underline{x}^0) \\ \dots \\ f_i(\underline{x}^0) \\ \dots \\ f_n(\underline{x}^0) \end{vmatrix} = \begin{vmatrix} f_1(\underline{x}) \\ \dots \\ f_i(\underline{x}) \\ \dots \\ f_n(\underline{x}) \end{vmatrix}$$

$${}^t \underline{J} \underline{\Delta x} + \underline{F}(\underline{x}^0) = \underline{F}(\underline{x})$$



$${}^t \underline{\underline{J}} \underline{\Delta x} + \underline{F}(\underline{x}^0) = \underline{F}(\underline{x})$$

$$\underline{F}(\underline{x}) = 0$$

$$\underline{x}^1 = \underline{x}^0 - {}^t \underline{\underline{J}}_{x^0}^{-1} \underline{F}(\underline{x}^0)$$

Iteration n :

$$\underline{x}^{n+1} = \underline{x}^n - {}^t \underline{\underline{J}}_{x^n}^{-1} \underline{F}(\underline{x}^n)$$



- At iteration k
 - Variables variations

$$\bullet \text{ Functions variation} \quad |x^{k+1} - x^k| \leq \epsilon * (1 + |x^k|)$$

$$|f(x^k)| \leq \epsilon$$

▪ Variables

Original problem

$$\begin{aligned} |f^{k+1}| &\leq \epsilon \\ |x^{k+1} - x^k| &\leq \epsilon * (1 + |x^k|) \end{aligned}$$

Scaling $F(X) = 0 \rightarrow \Phi(\Xi) = 0$

$$\Xi = Scale_x \cdot X$$

$$\Phi = Scale_f \cdot F(X)$$

All equations and variables are compared with the same reference value

New problem

$$\begin{aligned} |\chi_j^{k+1} - \chi_j^k| &\leq \epsilon * (1 + |\chi_j^k|) \quad \forall j \\ |\phi_j^{k+1}(\chi) - \phi_j^k(\chi)| &\leq \epsilon \quad \forall j \end{aligned}$$



- Forward numerical estimation

$$\frac{\partial f}{\partial x_i} = \frac{f(X + \Delta x_i) - f(X)}{\Delta x_i} \quad \forall i = 1, \dots, n_X + 1$$

- Computation time cost !
 - central derivatives avoided
 - Matrix Inversion
 - the matrix $J_{x^n}^{-1}$ is reused and not re-evaluated at each iteration
- Numerical noise (is the derivative a derivative ?)
 - $\Delta f(X) = |F(X)_0 - F(X)_{n_X+1}|$

- Initial guess
- Matrix inversion + derivatives
- Matrix inversion at each iteration $n \times n$ derivatives,
- Relaxation may be applied :

$$\tilde{\underline{x}}^{n+1} = \tilde{\underline{x}}^n - {}^t \underline{\underline{J}}_{\underline{x}^n}^{-1} \underline{F}(\tilde{\underline{x}}^n) \cdot (1 - q)$$

