

Yani's cheap plastic review of Newton-Raphson (NR) Method

Idea: Using linear approximation of the function $f(x)$ or set of functions $F(X)$ to solve them.

Taylor Series:

$$f(x) = f(x_0) + \sum_{k=1}^{\infty} \frac{f^k(x_0)}{k!} (x - x_0)^k = f(x_0) + f'(x_0) \cdot (x - x_0) + \frac{f''(x_0)}{2!} \cdot (x - x_0)^2 + \dots$$

Example: $f(x) = e^x$. Approximate e^1 and $e^{0.01}$ using Taylor Series expansion at $x_0 = 0$.

e^x is convenient since $f(x) = f'(x) = f''(x) = f^k(x) = e^x$ and they are all equal to 1 at $x = 0$ T.S. expansion gives:

$$e^x = e^0 + e^0(x - 0) + \frac{e^0}{2!}(x - 0)^2 + \dots = 1 + x + \frac{x^2}{2} + \dots$$

If we approximate e^x using only the first two terms: $e^x \approx 1 + x$

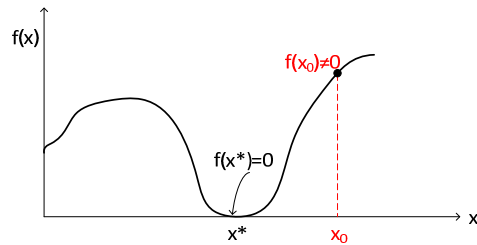
Let's calculate this for a couple of points:

x	e ^x (exact value)	Taylor Series Approximation
0.01	1.0105	1.01
1	2.718	2

i.e. Taylor Series approximation is more accurate for x closer to x_0

NR for scalar non-linear function

One equation and one unknown.



Start at x_0
Want to find Δx such that $\Delta x = x^* - x_0$

Taylor Series expansion around x_0 :

$$f(x) = f(x_0) + f'(x_0) \cdot (x - x_0) + H.O.T.$$

For $x = x^*$,

$$f(x^*) = f(x_0) + f'(x_0) \cdot (x^* - x_0) + H.O.T.$$

$$0 \approx f(x_0) + f'(x_0) \cdot \Delta x$$

Or,
$$\Delta x \approx -\frac{f(x_0)}{f'(x_0)}$$

$$\Delta \tilde{x} = -\frac{f(x_0)}{f'(x_0)}$$

- $\Delta \tilde{x}$ can approximate how much to change x_0 to get to x^* . We will call this the update term in NR.
- If x_0 and x^* are close, $\Delta \tilde{x}$ will be more accurate. (i.e. its value will be closer to the true Δx)

Steps: 1) Choose an initial value x_0 and set $k = 0$. (k keeps track of iteration #)

2) Solve for $\Delta \tilde{x}_k = -\frac{f(x_k)}{f'(x_k)}$

3) Let $x_{k+1} = x_k + \Delta \tilde{x}_k$

4) Check for convergence, i.e. $|f(x_{k+1})| < \varepsilon$ or $|\Delta \tilde{x}_k| < \varepsilon$

5) If converged, stop. Else, $k++$ and go to step 2.

NR for system of non-linear equations

$$F(X) = \begin{bmatrix} f_1(X) \\ f_2(X) \\ \vdots \\ f_n(X) \end{bmatrix} \quad \text{where} \quad X = \begin{bmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{bmatrix}$$

Taylor Series expansion around an (N-dimensional) point X^0 :
 $F(X) = F(X^0) + J(X^0) \cdot (X - X^0) + H.O.T.$

Jacobian: $n \times n$ matrix of partial derivatives:

$$J(X) = \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

Same as before,

$$F(X^*) = F(X^0) + J(X^0) \cdot (X^* - X^0) + H.O.T.$$

$$[0] \approx F(X^0) + J(X^0) \cdot (X^* - X^0)$$

$$J(X^0) \cdot \Delta \tilde{X} = -F(X^0)$$

Steps:

- Symbolic evaluation of the Jacobian $J(X)$ for a generic X

1) Choose an initial value X^0 and set $k = 0$.

2) Compute the Jacobian and the mismatch vector: Plug in X^k into $J(X)$ and $F(X)$

3) Solve the linear problem $J(X^k) \cdot \Delta \tilde{X}^k = -F(X^k)$ to find $\Delta \tilde{X}^k$

4) Let $X^{k+1} = X^k + \Delta \tilde{X}^k$

5) Check for convergence, i.e. $\|F(X^{k+1})\| < \varepsilon$ or $\|\Delta \tilde{X}^k\| < \varepsilon$

6) If converged, stop. Else, $k++$ and go to step 2.