

## Newton Raphson 2D $\wedge$ Non-Linear:

Linear: 
$$@x_i, y_i \rightarrow \begin{bmatrix} \partial f / \partial x & \partial f / \partial y \\ \partial g / \partial x & \partial g / \partial y \end{bmatrix} \begin{bmatrix} \Delta x \\ \Delta y \end{bmatrix} = \begin{bmatrix} -f(x_i, y_i) \\ -g(x_i, y_i) \end{bmatrix}$$

use Cramer's Rule.

## Non-Linear Least Squares:

• error:  $\left| \frac{z_{i+1} - z_i}{z_{i+1}} \right| < \text{error}$   
add all Params' error

$R_i = \text{Resid}$ ,  $J_i = \text{Jacobian w/r param}$

$B_i = \Delta \text{Parameters' value}$

$$B_i = (J_i^T J_i)^{-1} J_i^T R_i$$

$$R_{ij} = y_j - f(\text{Params})$$

• Solving for param change.

Univariate: 2D Golden Search where each var,  $x, y, z$ , is checked one at a time.

• checking  $x$  means we use  $f(x_1, y_0, z_0) \wedge f(x_2, y_0, z_0)$

## Steepest Ascent:

$$\nabla f = \langle g_{Fx}, g_{Fy} \rangle = \left\langle \frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right\rangle$$

$$x_{i+1} = x_i + k^* \left( \frac{\partial f}{\partial x} \Big|_{x_i, y_i} \right)$$

$$y_{i+1} = y_i + k^* \left( \frac{\partial f}{\partial y} \Big|_{x_i, y_i} \right)$$

$$k^* = \text{goldsearch}(x_i, y_i)$$

$$\rightarrow k_{\text{low}} = \left[ \frac{(x_L - x_i)}{\frac{\partial f}{\partial x} \Big|_{x_i, y_i}}, \frac{(y_L - y_i)}{\frac{\partial f}{\partial y} \Big|_{x_i, y_i}} \right]$$

$$\rightarrow k_{\text{upp}} = \left[ \frac{(x_u - x_i)}{\frac{\partial f}{\partial x} \Big|_{x_i, y_i}}, \frac{(y_u - y_i)}{\frac{\partial f}{\partial y} \Big|_{x_i, y_i}} \right]$$

$$\rightarrow l = \min(\text{abs}(k_{\text{low}})), l \in \mathcal{I}$$

$$\rightarrow u = \min(\text{abs}(k_{\text{upp}})), u \in \mathcal{I}$$

$$\rightarrow k_u = k_{\text{low}}(u); k_l = k_{\text{upp}}(u)$$

$$\rightarrow \text{if } k_l > k_u \quad \# \text{ checks } F(x(k), y(k))$$
$$\rightarrow k_t = k_l$$
$$\rightarrow k_l = k_u$$
$$\rightarrow k_u = k_t$$

$\rightarrow k^*$  is a parameterization like  $t$

## Golden Search: $d = \frac{1}{2}(\sqrt{5}-1)$

$$x \in [x_L, x_u]$$

$$1.) x_1 = x_L + d(x_u - x_L)$$

$$x_2 = x_u - d(x_u - x_L)$$

$$2.) \text{Evaluate } f(x_1) \text{ \& } f(x_2)$$

$$3.) \cdot f(x_1) > f(x_2)$$

• erase  $[x_L, x_1]$

$$\rightarrow x_1 = x_L + d(x_u - x_L)$$

$\rightarrow$  Eval new  $f(x_1)$ , iterate

$$\cdot f(x_1) < f(x_2)$$

• erase  $[x_1, x_u]$

$$\rightarrow x_u = x_1$$

$$\rightarrow x_1 = x_2$$

$$\rightarrow x_1 = x_u - d(x_u - x_L)$$

$\rightarrow$  Eval  $f(x_2)$ , iterate...

$$3.) \cdot J(x_1) > J(x_2)$$

• erase  $[x_L, x_2]$

$$\rightarrow x_L = x_2; x_2 = x_1$$

$$\cdot x_1 = x_u \quad \text{if } (x_u = x_L)$$

$\rightarrow$  Eval  $f(x_2)$ , iterate...