

NEWTON - RAPHSON

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

⋮

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

TAYLOR SERIES

$$f_i(x_1, \dots, x_n) = f_i(x_1^*, \dots, x_n^*)$$

$$+ (x_1 - x_1^*) \left. \frac{\partial f_i}{\partial x_1} \right|_{\vec{x}^*}$$

$$+ (x_2 - x_2^*) \left. \frac{\partial f_i}{\partial x_2} \right|_{\vec{x}^*}$$

⋮

$$+ (x_n - x_n^*) \left. \frac{\partial f_i}{\partial x_n} \right|_{\vec{x}^*}$$

$$\underbrace{f_i(x_1, \dots, x_n)}_{(2)} = f_i(x_1^*, \dots, x_n^*) + \underbrace{\sum_{j=1}^n (x_j - x_j^*) \frac{\partial f_i}{\partial x_j}}_{(1)} \bigg|_{\vec{x}^*} + \dots$$

$$\begin{aligned} \textcircled{1} \quad \sum_{j=1}^n \frac{\partial f_i}{\partial x_j} (x_j - x_j^*) &= \begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix} \begin{bmatrix} x_1 - x_1^* \\ \vdots \\ x_n - x_n^* \end{bmatrix} \\ &\quad \underbrace{\hspace{10em}}_{\text{JACOBIAN MATRIX}} \end{aligned}$$

$$\textcircled{2} \quad \begin{pmatrix} f_1(\vec{x}) \\ f_2(\vec{x}) \\ \vdots \\ f_n(\vec{x}) \end{pmatrix}$$

$$\vec{f}(\vec{x}) = \vec{f}(\vec{x}^*) + \bar{J} \cdot (\vec{x} - \vec{x}^*)$$

where $\vec{x} = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$

$$J_{ij} = \left. \frac{\partial f_i}{\partial x_j} \right|_{\vec{x}^*}$$

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

• IF $\vec{\Delta x} = (\vec{x}^* - \vec{x})$ is step toward solution

• IF \vec{x} is solution: LHS = 0

$$\vec{f}(\vec{x}^*) = \bar{J} \cdot \vec{\Delta x}$$

$$\vec{b} = \bar{A} \vec{x}$$

$$\vec{\Delta x} = ?$$

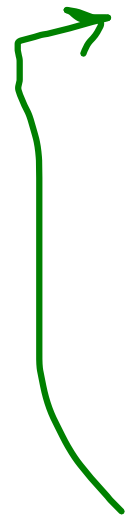
$$= \bar{J}^{-1} \cdot \vec{f}(\vec{x}^*) = \Delta x$$

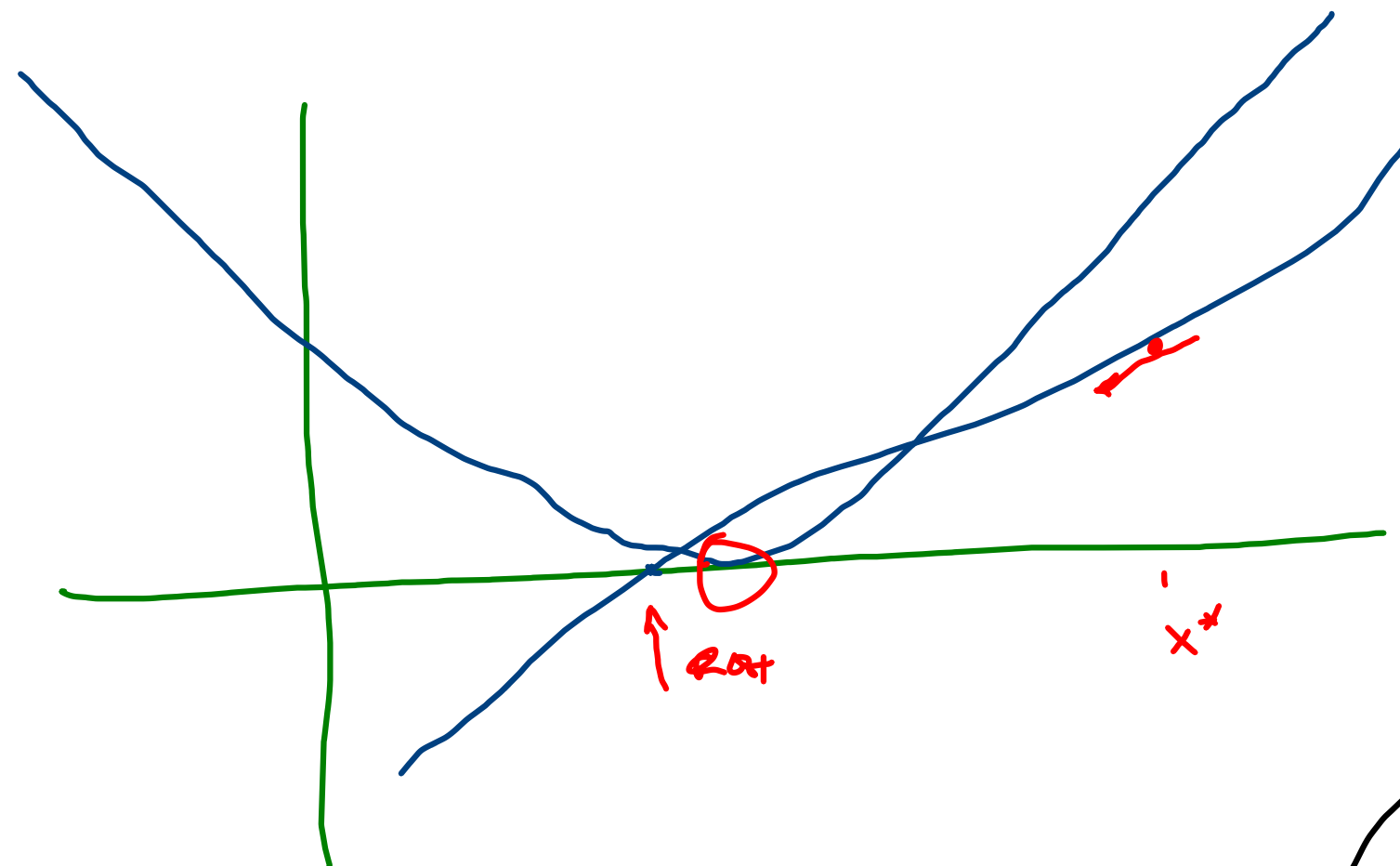
⑤ INITIAL GUESS \vec{x}^*

① CALCULATE $\vec{f}(\vec{x}^*)$ AND $\bar{J}|_{\vec{x}^*}$

② SOLVE $\bar{J} \Delta \vec{x} = -\vec{f}$ FOR $\Delta \vec{x}$

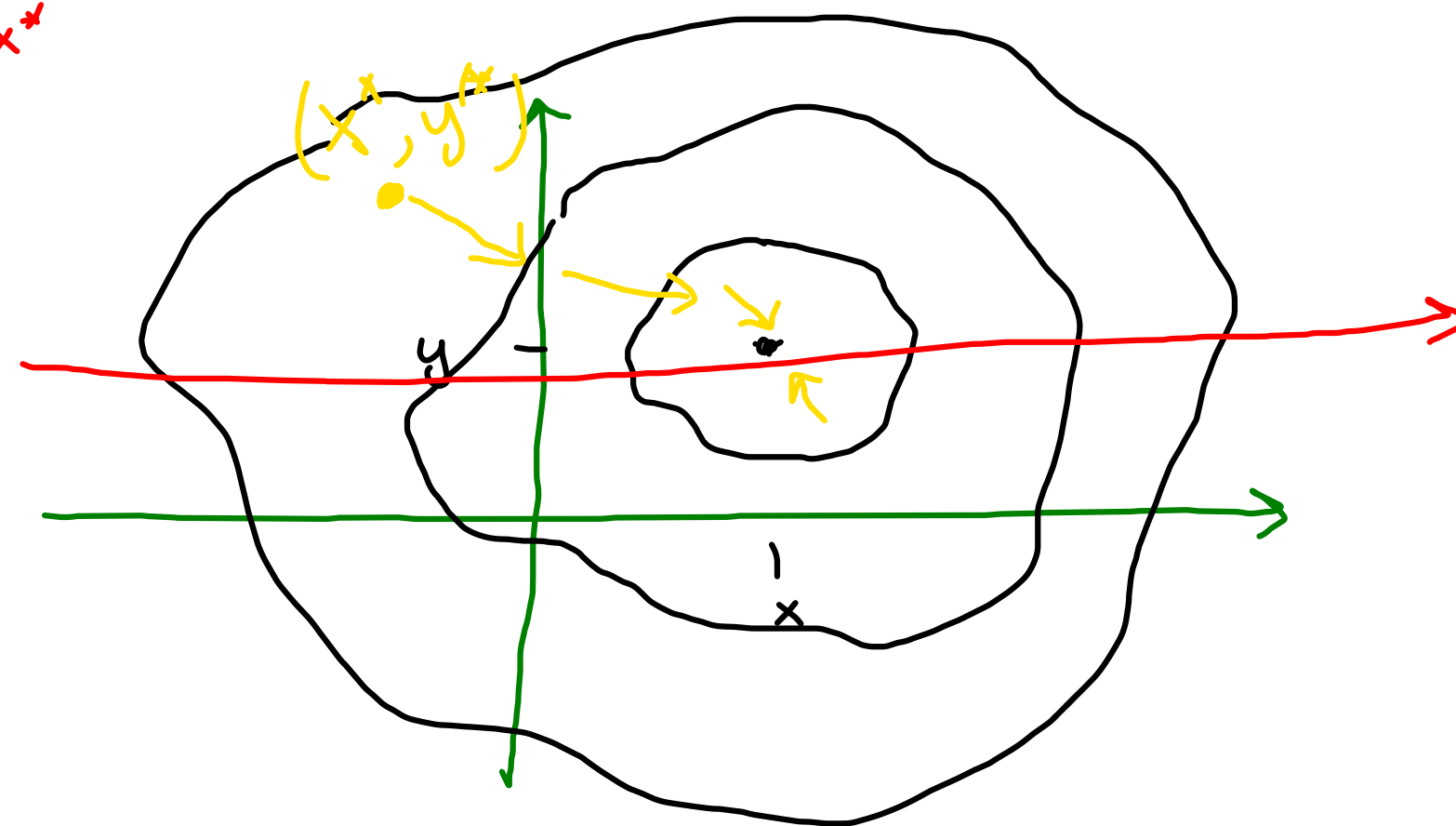
③ $\vec{x}' = \vec{x}^* - \Delta \vec{x}$





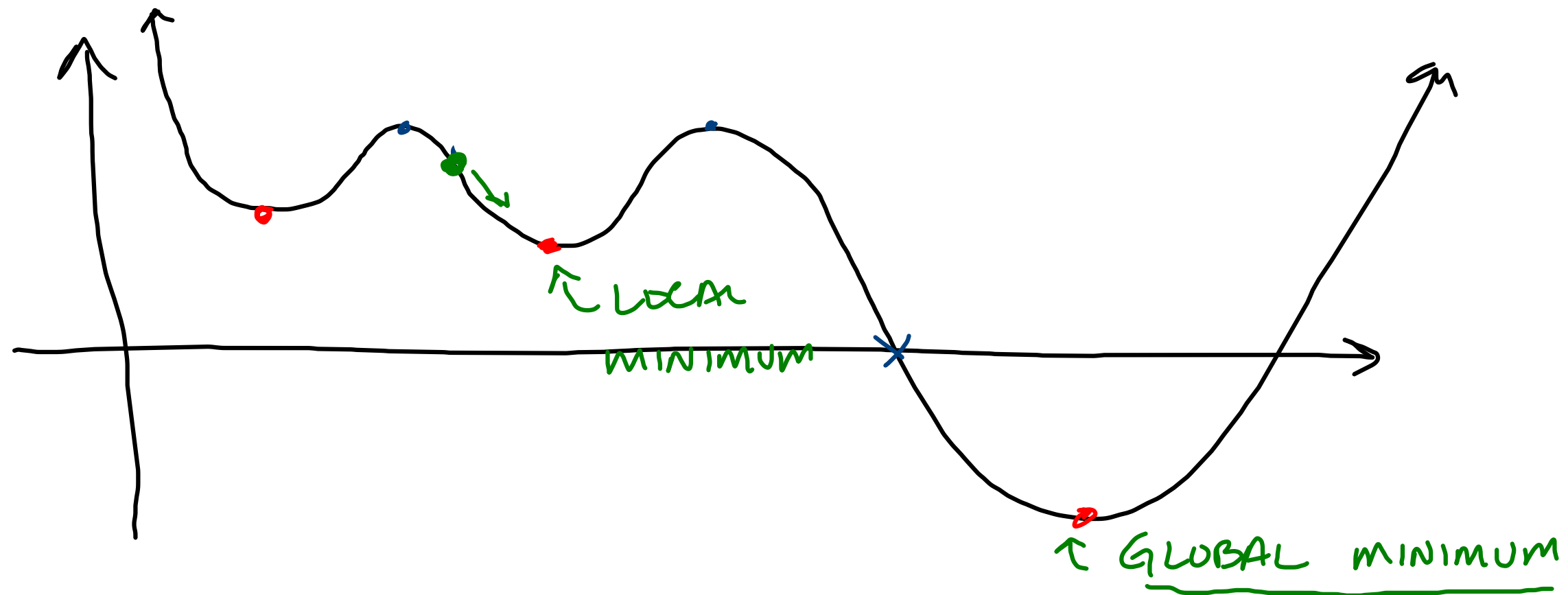
$$\Delta x = - \frac{f(x^*)}{f'(x^*)}$$

EXTREMUM OF FUNCTION
(OPTIMIZATION)



IF \vec{x} SATISFIES $\vec{f}(\vec{x}) = 0$

\vec{x} IS AN EXTREMUM OF $\vec{g}(\vec{x})$ IFF $\vec{\nabla}_{\vec{x}} \vec{g} = \vec{f}$



① GOOD GUESS REQUIRED