

COL726: Eigenvalues / Nonlinear equations

$$\underline{Q}^{(0)} \longrightarrow \underline{Q}^{(1)} \longrightarrow \underline{Q}^{(2)} \longrightarrow \dots \xrightarrow{\text{lim}} Q$$

Simultaneous iteration

$$\begin{array}{c} \downarrow \quad \downarrow \quad \downarrow \\ A^{(0)} \longrightarrow A^{(1)} \longrightarrow A^{(2)} \longrightarrow \dots \xrightarrow{\text{lim}} \Lambda \end{array}$$

QR algorithm

$$\curvearrowright = \underline{Q}^{(0)\top} A \underline{Q}^{(0)}$$

Conditioning of eigenvalues

$$A = \underline{V} \underline{\Lambda} \underline{V}^{-1}$$

μ is eigenvalue of perturbed matrix $A + \delta A$

$$\underline{V}^{-1} (A + \delta A) \underline{V} = \underbrace{\underline{V}^{-1} A \underline{V}}_A + \underbrace{\underline{V}^{-1} \delta A \underline{V}}_{\delta \Lambda} \quad \mu \text{ is an eigenvalue of } \Lambda + \delta \Lambda$$

$$\Lambda \rightarrow \lambda_1, \lambda_2, \dots, \lambda_j, \dots, \lambda_m$$

$$\Lambda + \underline{\underline{\delta\Lambda}} \rightarrow$$

$$\mu$$

$$\Rightarrow |\mu - \lambda_j| \leq \underline{\underline{\|\delta\Lambda\|_2}}$$

$$\exists \vec{v} : (\Lambda + \delta\Lambda) \vec{v} = \mu \vec{v}$$

$$\underbrace{\sigma_{\min}(\mu I - \Lambda)}_{\downarrow} \|\vec{v}\|_2 \leq \|(\mu I - \Lambda) \vec{v}\|_2 = \|\delta\Lambda \vec{v}\|_2 \leq \underline{\underline{\|\delta\Lambda\|_2}} \|\vec{v}\|_2$$

$$\min_{j=1 \dots m} |\mu - \lambda_j|$$

$$A = V \Lambda V^{-1}, \quad \delta\Lambda = V^{-1} \delta A V$$

$$|\mu - \lambda_j| \leq \|\delta\Lambda\|_2 \approx \underbrace{\|V^{-1}\|} \|\delta A\| \underbrace{\|V\|}$$

$$\xrightarrow{+0.01} \lambda = 1.5 \pm 0.86i$$

$$A = \begin{bmatrix} 101 & -110 \\ 90 & -98 \end{bmatrix}$$

$$\lambda = 2, 1$$

$$\kappa(V) \approx 402$$

$$= \kappa(V) \|\delta A\|$$

$$\text{But if } A^T A = A A^T \Rightarrow \kappa(V) = 1$$

Nonlinear equations

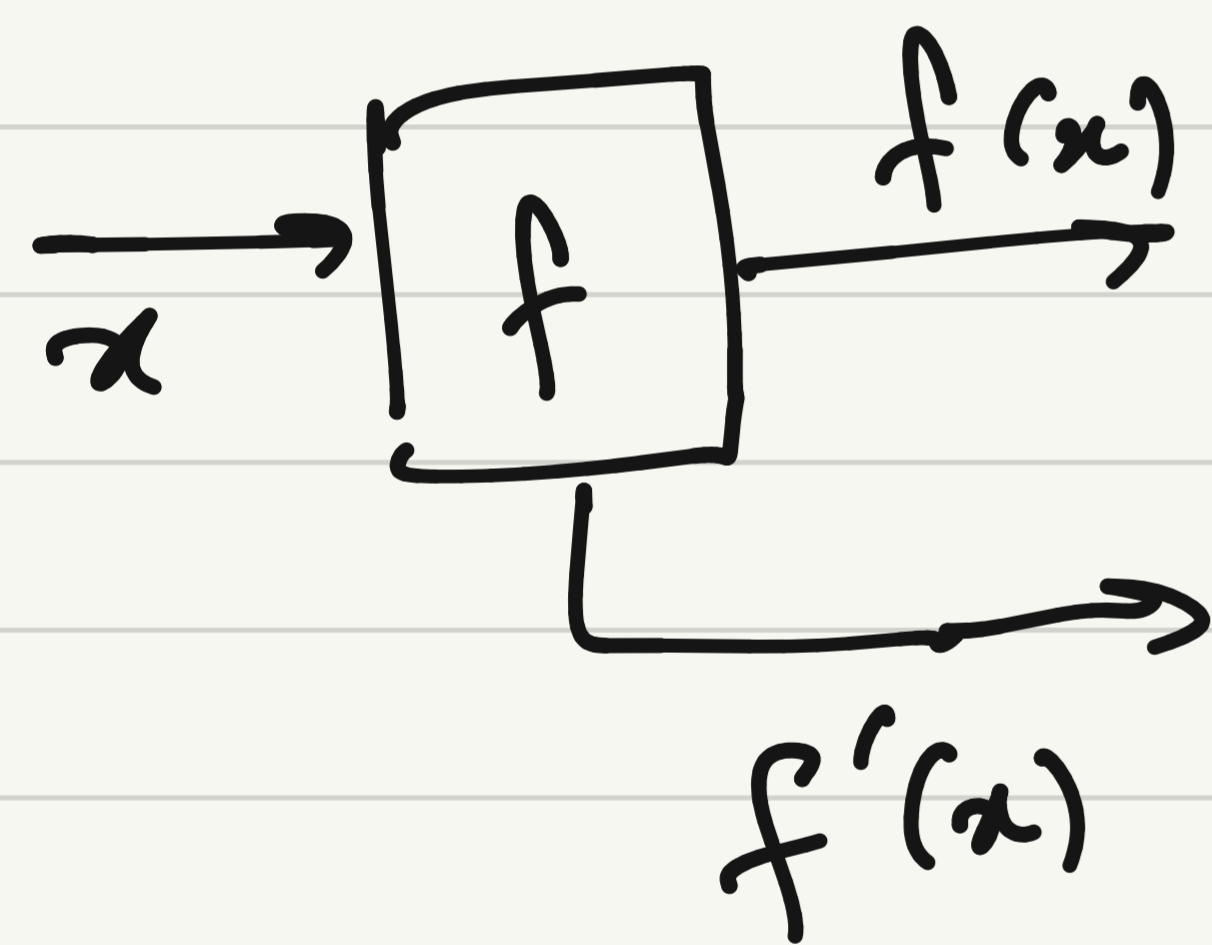
$$f: \mathbb{R} \rightarrow \mathbb{R}$$

find $x: f(x) = 0$: root finding

$$\underline{f: \mathbb{R}^n \rightarrow \mathbb{R}^n}$$

x is a root of f

$$f(x) = y \iff f(x) - y = 0$$



Existence & uniqueness

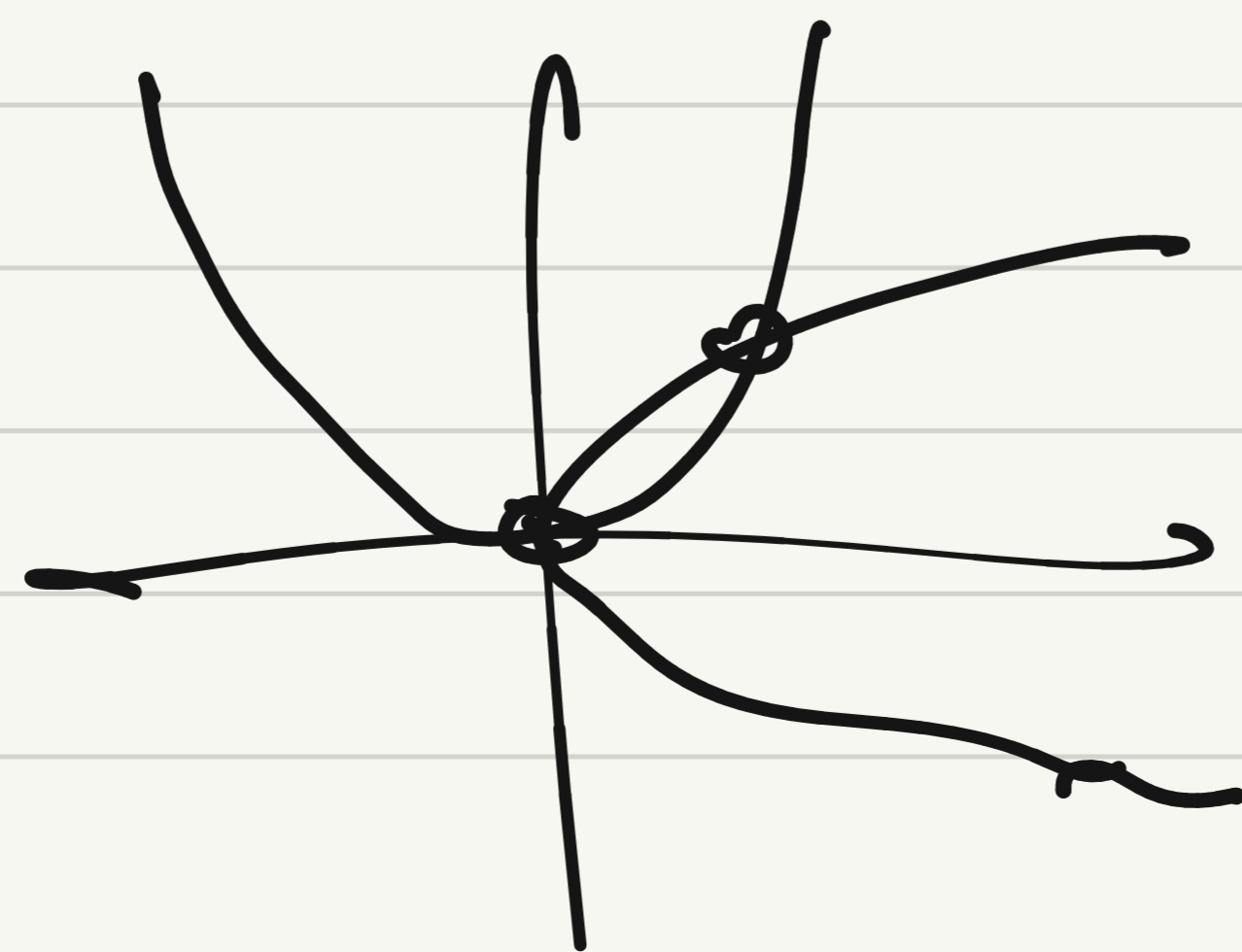
Only consider equal # eqs. & vars

$$\vec{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$f(\vec{x}) = \begin{bmatrix} q_1(x_1, x_2) \\ q_2(x_1, x_2) \end{bmatrix}$$

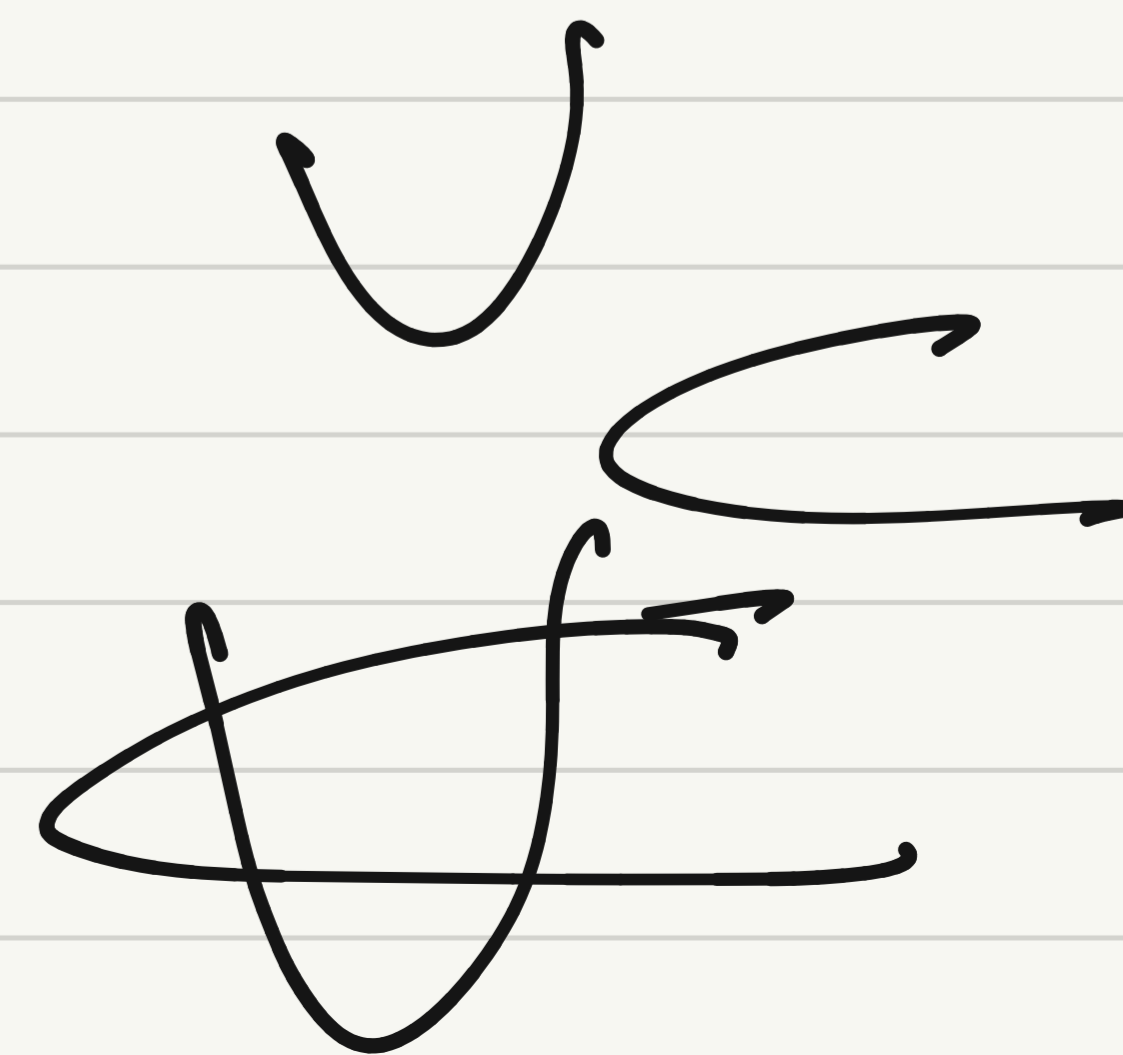
$$q_1(x_1, x_2) = 0$$

$$q_2(x_1, x_2) = 0$$



$$c_1 + x_1^2 = x_2$$

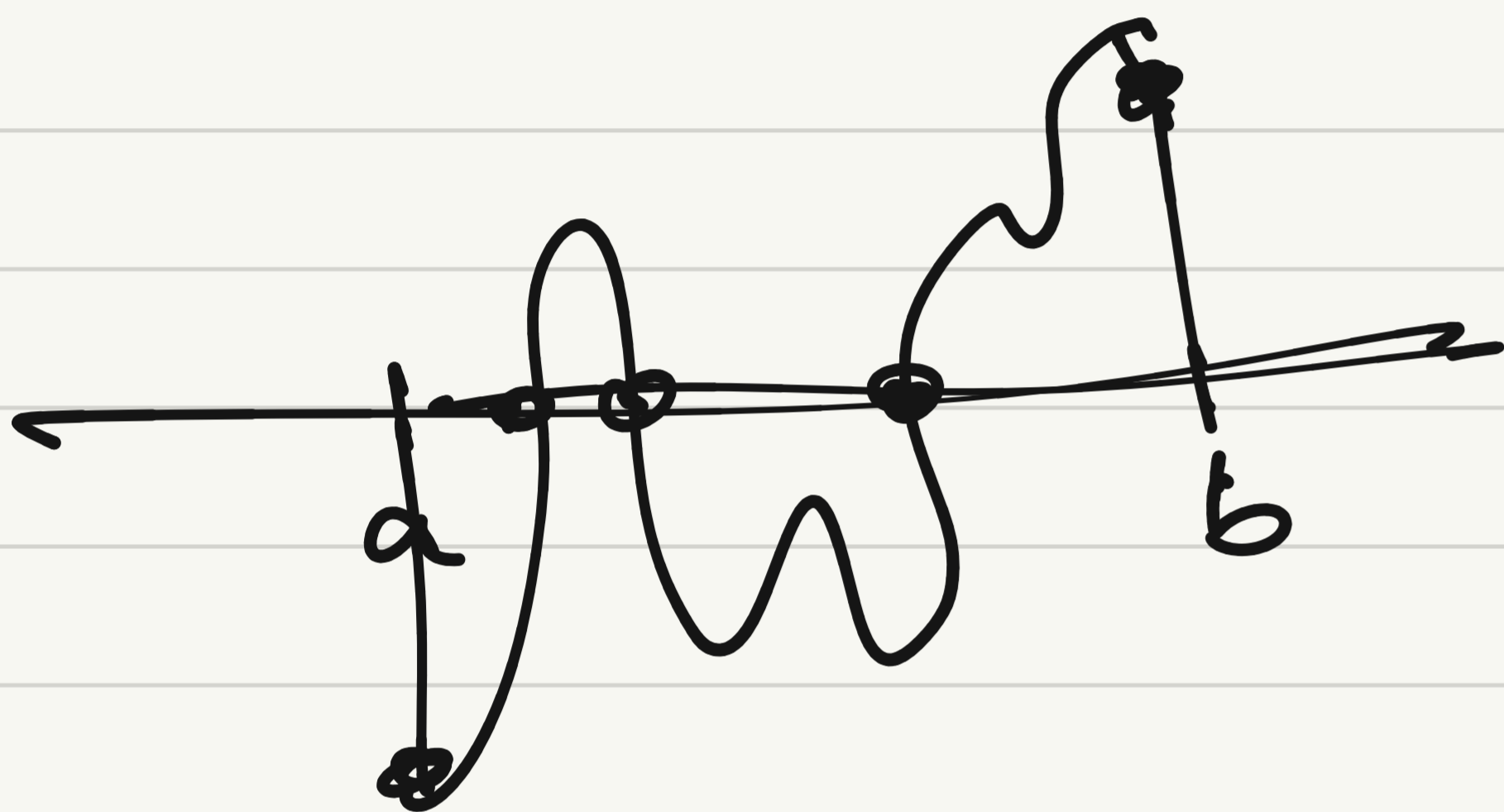
$$c_2 + x_2^2 = x_1$$



Local guarantees of existence & uniqueness

Intermediate value theorem: $f: \mathbb{R} \rightarrow \mathbb{R}$ continuous,

If $f(a) < 0$, $f(b) > 0$ then $\exists x \in [a, b]$ s.t. $f(x) = 0$



Interval $[a, b]$ with $f(a)$, $f(b)$ having diff. signs

Signs : Bracket

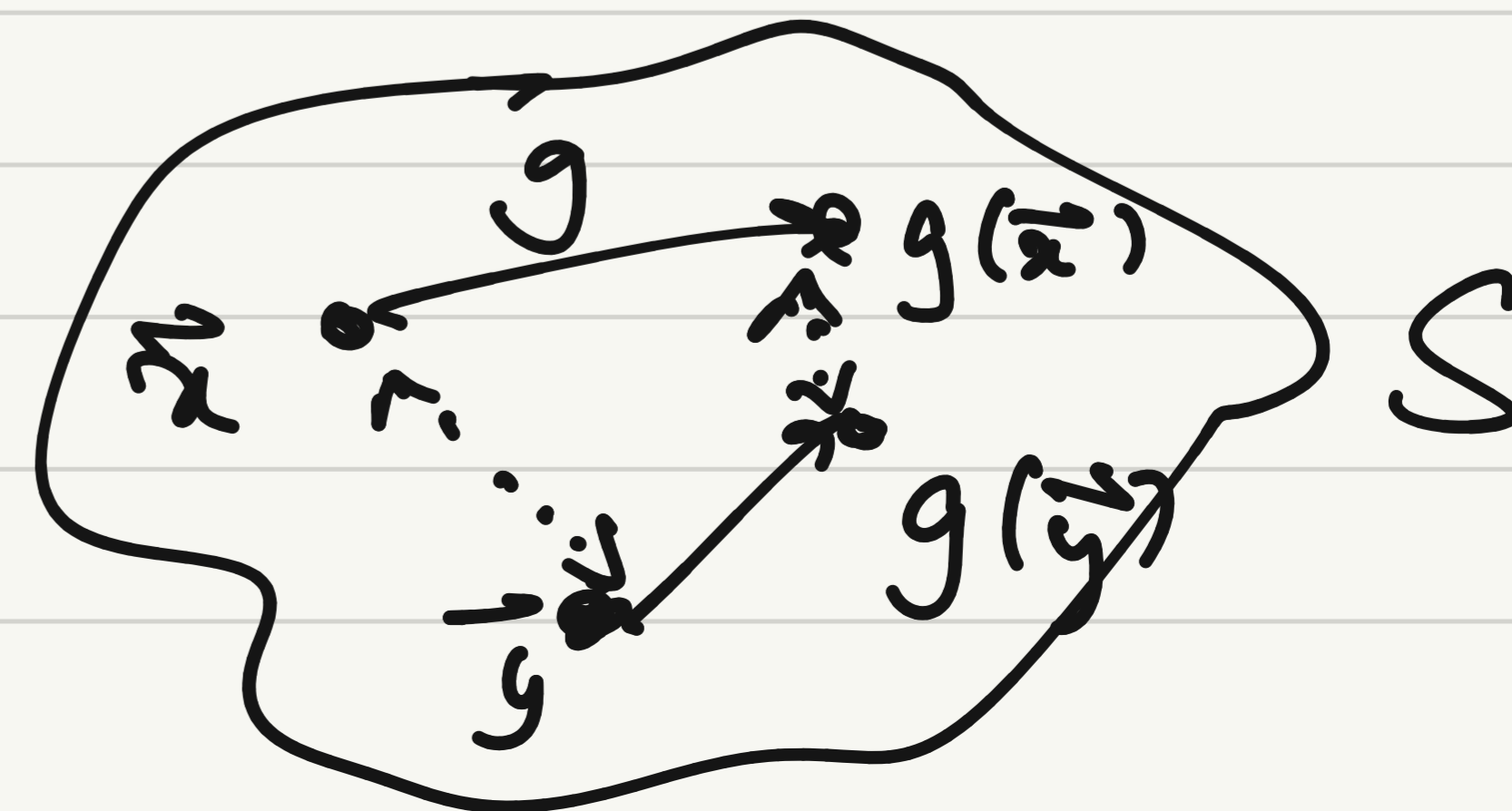
Contraction mapping thm. \therefore

If $g: S \rightarrow S$ is such that

$$\|g(\vec{x}) - g(\vec{y})\| < \|\vec{x} - \vec{y}\| \quad \forall \vec{x}, \vec{y}$$

then g is a Contraction

$$S \subseteq \mathbb{R}^n$$



If g is contraction on compact set S then it has unique point $\vec{x} \in S$

s.t. $g(\vec{x}) = \vec{x}$: fixed point of g

Let $f(\vec{x}) = \vec{x} - g(\vec{x})$: f has unique solution!

Inverse function Thm.

Jacobian

of $f: \mathbb{R}^n \rightarrow \mathbb{R}^n$

If Jacobian is nonsingular

then f is locally invertible

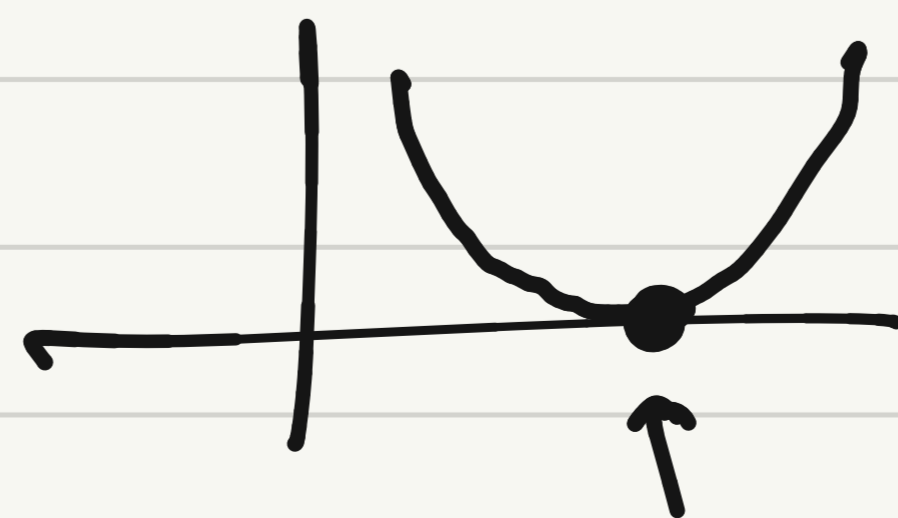
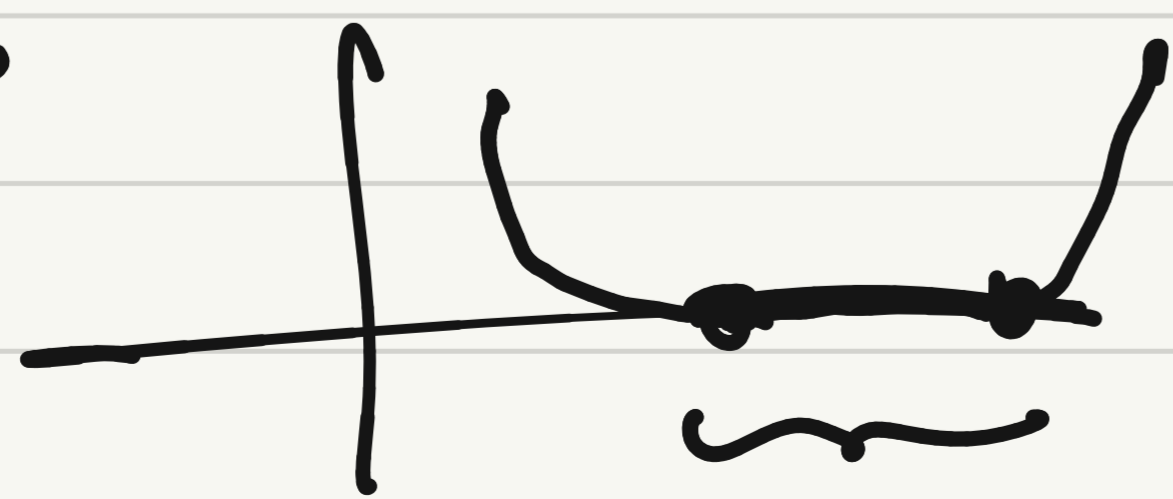
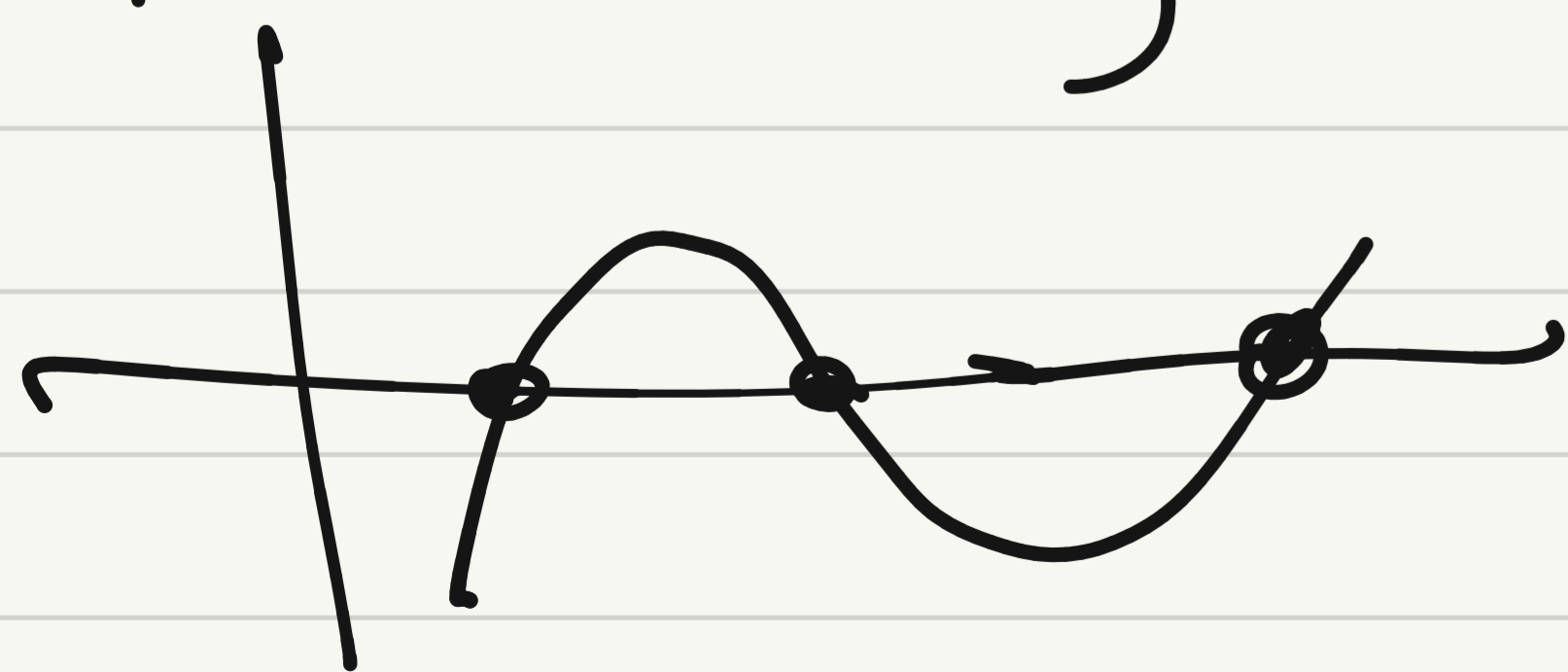
\Rightarrow solution \vec{x}_* is locally unique

$J(\vec{x}) =$

$$\begin{bmatrix} \frac{\partial f_1}{\partial x_1} & \frac{\partial f_1}{\partial x_2} & \dots & \frac{\partial f_1}{\partial x_n} \\ \frac{\partial f_2}{\partial x_1} & \frac{\partial f_2}{\partial x_2} & \dots & \vdots \\ \vdots & \vdots & \dots & \vdots \\ \frac{\partial f_n}{\partial x_1} & \dots & \dots & \frac{\partial f_n}{\partial x_n} \end{bmatrix}$$

If not, we say solution is degenerate

$$f(\vec{x} + \delta \vec{x}) = f(\vec{x}) + J(\vec{x}) \cdot \delta \vec{x} + o(\|\delta \vec{x}\|^2)$$



If $f(x) = 0, f'(x) = 0, f''(x) = 0, \dots, f^{(n-1)}(x) = 0$
 $f^{(n)} \neq 0$ } \Rightarrow multiplicity of root is n .

Conditioning:

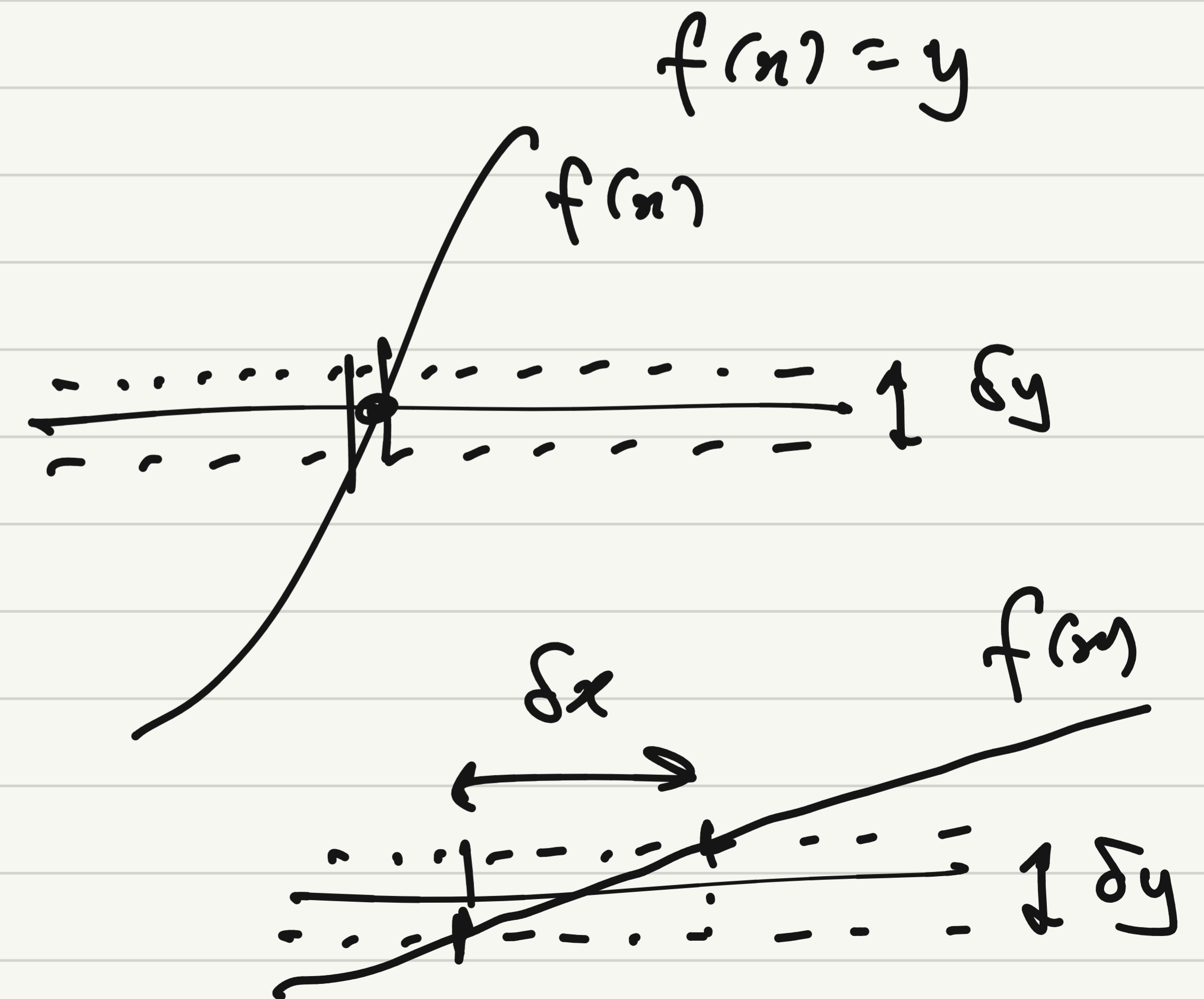
$$f(x_*) = 0, \rightarrow f(x) = y$$

$$f: \mathbb{R} \rightarrow \mathbb{R} \Rightarrow x_* = f^{-1}(y)$$

$$\text{Abs. cond. num} = \frac{|\delta x|}{|\delta y|} = \left| \frac{1}{f'(x_*)} \right|$$

$$\text{In } \mathbb{R}^n, \text{ abs. cond. num.} = \|J^{-1}(x_*)\|$$

$$f^{-1} = g, \quad g'(y) = \frac{1}{f'(x)}$$



Convergence

$$x_0 \rightarrow x_1 \rightarrow x_2 \rightarrow \dots \rightarrow x_k \rightarrow \dots \xrightarrow{\text{lim}} x_*$$

flops

error: $e_k = x_k - x_*$

residual: $f(x_k)$

Rate of convergence

$$\lim_{k \rightarrow \infty} \frac{\|e_{k+1}\|}{\|e_k\|} = C < 1$$

: convergence is linear with rate const. C

gain $\sim -\log_{10} C$ decimal digits per iter

eg. power iter: $C = |\lambda_2|/|\lambda_1|$

$$\lim_{k \rightarrow \infty} \frac{\|e_{k+1}\|}{\|e_k\|^r} = C \quad (\text{not necessarily } < 1): \text{ convergence is } \boxed{\text{super linear}} \text{ if } r > 1$$

$r = 2$: quadratic, $r = 3$: cubic

1D algorithms : $f: \mathbb{R} \rightarrow \mathbb{R}$

- bisection method
- fixed-point iterations
- Newton's method
- Secant method

Bisection : input $[a, b]$ s.t. $\text{sign}(f(a)) \neq \text{sign}(f(b))$

while $b - a > \text{tolerance}$:

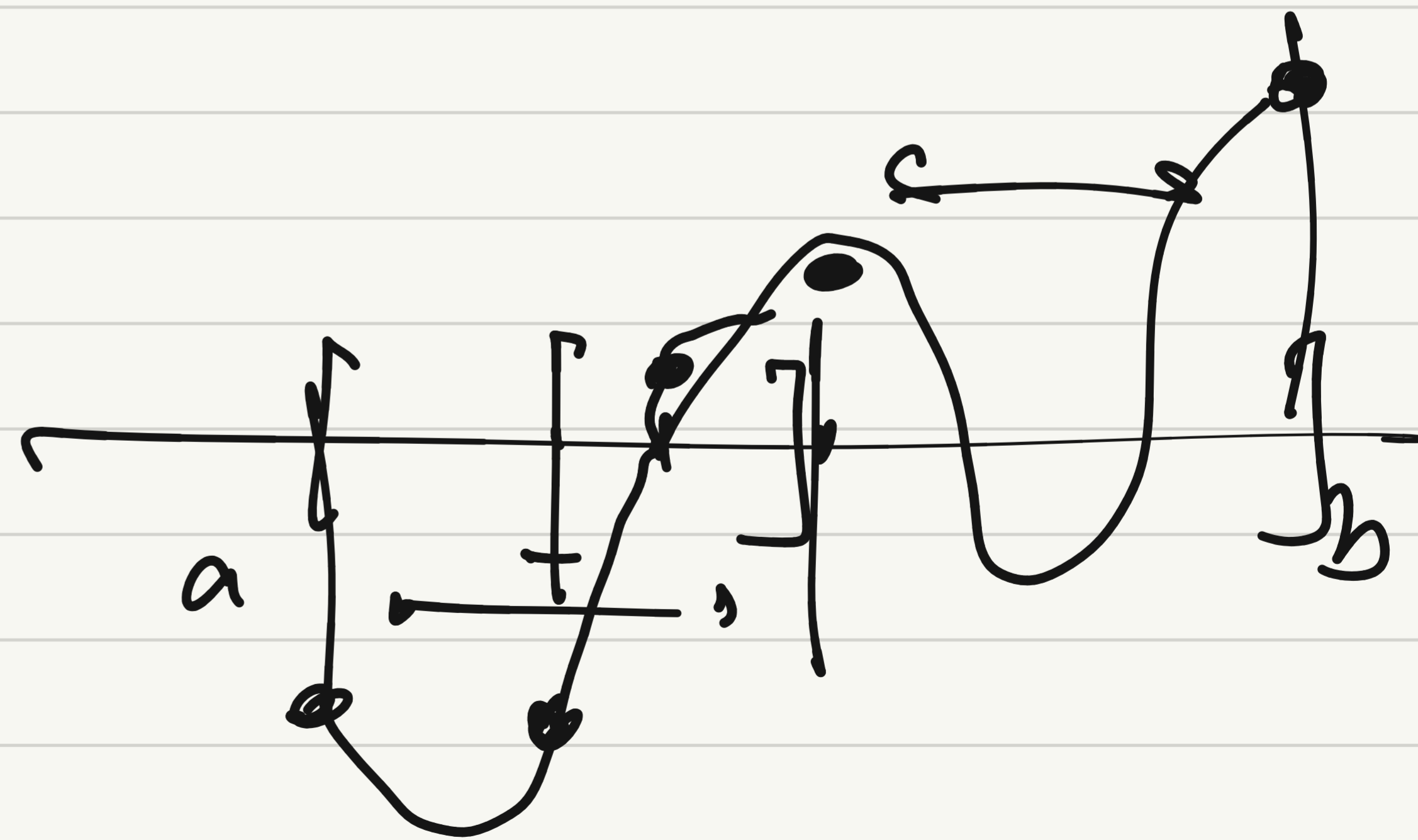
$$m = a + \frac{(b-a)}{2}$$

if $\text{sign}(f(m)) = \text{sign}(f(a))$

$a = m$

else

$b = m$



No x_k , $[a_k, b_k]$ instead

$$\text{Error } e_k = \underbrace{b_k - a_k}$$

$$\frac{|e_{k+1}|}{|e_k|} = \frac{1}{2}$$

fixed point iteration

$f(x) = 0$: find $g: \mathbb{R} \rightarrow \mathbb{R}$ s.t. $g(x) = x$ iff $f(x) = 0$

$$x_0 \rightarrow x_1 = g(x_0) \rightarrow x_2 = g(x_1) \rightarrow \dots$$

$$f(x) = \underbrace{x^2 - x - 2 = 0}$$

$$\rightarrow x = x^2 - 2 = g(x)$$

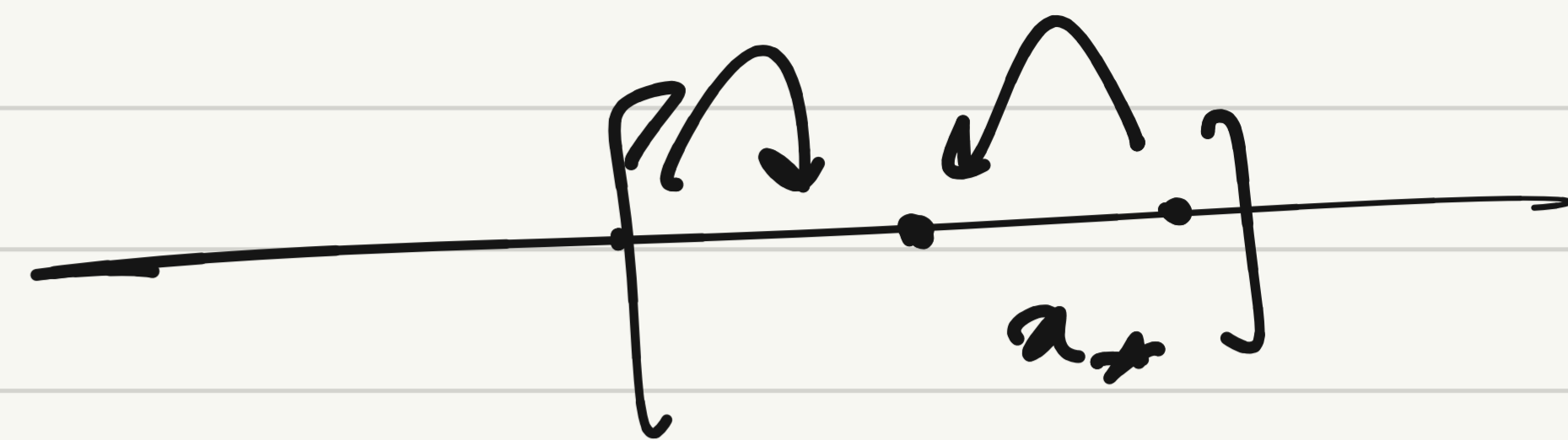
$$\rightarrow x - 1 - 2/x = 0 \Rightarrow x = 1 + \frac{2}{x}$$

$$\rightarrow x^2 = x + 2 \Rightarrow x = \sqrt{x + 2} = g(x)$$

$\rightarrow \dots$

$$x_0 \rightarrow \dots \rightarrow x_k \rightarrow x_{k+1} = g(x_k) \rightarrow \dots$$

When does fixed point iteration converge?



If there is some neighborhood S containing x_*

s.t. g is contraction mapping S to itself,

and $x_0 \in S$ then $|e_k| = |x_k - x_*|$ decreases every iteration

$$|g(x_k) - g(x_*)| \leq |x_k - x_*|$$