

**Definition 1 (Simple Iteration).** Suppose that  $g$  is a real-valued function, defined and continuous on a bounded closed interval  $[a, b]$  of the real line, and let  $g(x) \in [a, b]$  for all  $x \in [a, b]$ . Given that  $x_0 \in [a, b]$ , the recursion defined by

$$x_{k+1} = g(x_k) \quad (1)$$

is called simple iteration; the numbers  $x_k$ ,  $k \geq 0$ , are referred to as iterates.

**Definition 2 (Contraction).** Let  $g$  be a real-valued function, defined and continuous on a bounded closed interval  $[a, b]$  of the real line. Then,  $g$  is said to be a contraction on  $[a, b]$  if there exists a constant  $L$  such that  $0 < L < 1$  and

$$|g(x) - g(y)| \leq L|x - y| \quad \forall x, y \in [a, b] \quad (2)$$

**Definition 3 (Stable, Unstable Fixed Point).** Suppose that  $g$  is a real-valued function, defined and continuous on a bounded closed interval  $[a, b]$  of the real line, and let  $g(x) \in [a, b]$  for all  $x \in [a, b]$ , and let  $\xi$  denote a fixed point of  $g$ .  $\xi$  is a stable fixed point of  $g$  if the sequence  $(x_k)$  defined by the iteration  $x_{k+1} = g(x_k)$ ,  $k \geq 0$ , converges to  $\xi$  whenever the starting value  $x_0$  is sufficiently close to  $\xi$ . Conversely, if no sequence  $(x_k)$  defined by this iteration converges to  $\xi$  for any starting value  $x_0$  close to  $\xi$ , except for  $x_0 = \xi$ , then we say that  $\xi$  is an unstable fixed point of  $g$ .

**Definition 4 (Rate of Convergence).** Suppose  $\xi = \lim_{k \rightarrow \infty} x_k$ . Define  $E_k = |x_k - \xi|$ .

**Definition 5 (Newton's Method).** Newton's method for the solution of  $f(x) = 0$  is defined by

$$x_{k+1} = x_k - \frac{f(x_k)}{f'(x_k)} \quad (3)$$

**Definition 6 (Secant Method).** The secant method is defined by

$$x_{k+1} = x_k - f(x_k) \frac{x_k - x_{k-1}}{f(x_k) - f(x_{k-1})} \quad (4)$$

**Definition 7. (Unitary Matrix)** A matrix  $Q = [q_1 \dots q_n] \in \mathbb{R}^{m \times n}$  is unitary if and only if  $\langle q_i, q_j \rangle = \delta_{ij}$ .

**Definition 8. (Norm)** Suppose that  $\mathcal{V}$  is a linear space over the field  $\mathbb{R}$ . The *nonnegative* real-valued function  $\|\cdot\|$  is a norm on  $\mathcal{V}$  if the following axioms are satisfied: Fix  $v \in \mathcal{V}$

1. Positivity:  $\|v\| = 0$  if and only if  $v = 0$
2. Scale Preservation:  $\|\alpha v\| = |\alpha|\|v\|$  for all  $\alpha \in \mathbb{R}$
3. Triangle Inequality:  $\|v + w\| \leq \|v\| + \|w\|$ .

**Definition 9 (Operator Norm).** Let  $A$  be an  $m \times n$  matrix. That is,  $A$  is a linear transformation from  $\mathbb{R}^n$  to  $\mathbb{R}^m$ . Then the operator norm (or subordinate matrix norm) of  $A$  is

$$\|A\|_{p,q} = \sup_{x \in \mathbb{R}^n, x \neq 0} \frac{\|Ax\|_q}{\|x\|_p}. \quad (5)$$

**Definition 10 (Absolute Condition Number).**

$$Cond(f) = \sup_{x, y \in D, x \neq y} \frac{\|f(x) - f(y)\|}{\|x - y\|} \quad (6)$$

**Definition 11 (Absolute Local Condition Number).**

$$Cond_x(f) = \sup_{x + \delta x \in D, \delta x \neq 0} \frac{\|f(x + \delta x) - f(x)\|}{\|\delta x\|} \quad (7)$$

**Definition 12 (Relative Local Condition Number).**

$$cond_x(f) = \sup_{x + \delta x \in D, \delta x \neq 0} \frac{\|f(x + \delta x) - f(x)\| / \|f(x)\|}{\|\delta x\| / \|x\|} \quad (8)$$

**Definition 13 (Condition Number of a Nonsingular Matrix).** The condition number of a nonsingular matrix  $A$  is defined by

$$\kappa(A) = \|A\| \|A^{-1}\| \quad (9)$$

If  $\kappa(A) \gg 1$ , the matrix is said to be ill-conditioned.

**Definition 14 (Symmetric, Positive Definite, spd).** The real matrix  $A$  is said to be symmetric if  $A = A^T$ . A square  $n \times n$  matrix is called positive definite if

$$\mathbf{x}^T A \mathbf{x} > 0 \quad (10)$$

for all  $\mathbf{x} \in \mathbb{R}^n$ ,  $\mathbf{x} \neq 0$ .

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**Algorithm 1** Cholesky Factorization

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**Require:**  $A \in \mathbb{R}^{n \times n}$ , SPD

$L_1 \leftarrow \sqrt{a_{11}}$   
**for**  $k \leftarrow 2, 3, \dots, n$  **do**  
    Solve  $L_{k-1}l_k = a_k$  for  $l_k$   
     $l_{kk} \leftarrow \sqrt{a_{kk} - l_k^T l_k}$   
     $L_k \leftarrow \begin{pmatrix} L_{k-1} & 0 \\ l_k^T & l_{kk} \end{pmatrix}$   
**end for**

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**Definition 15 (Cauchy Sequence).** A sequence  $(\mathbf{x}^{(k)}) \subset \mathbb{R}^n$  is called a Cauchy sequence in  $\mathbb{R}^n$  if for any  $\epsilon > 0$  there exists a positive integer  $k_0 = k_0(\epsilon)$  such that

$$\|\mathbf{x}^{(k)} - \mathbf{x}^{(m)}\|_\infty < \epsilon \quad \forall k, m \geq k_0(\epsilon) \quad (11)$$

**Definition 16 (Continuous function).** Let  $D \subset \mathbb{R}^n$  be nonempty and  $f : D \rightarrow \mathbb{R}^n$ . Given  $\boldsymbol{\xi} \in D$ ,  $f$  is continuous at  $\boldsymbol{\xi}$  if for every  $\epsilon > 0$ , there exists a  $\delta = \delta(\epsilon) > 0$  such that for every  $\mathbf{x} \in B(\boldsymbol{\xi}; \delta) \cap D$

$$\|f(\mathbf{x}) - f(\boldsymbol{\xi})\|_\infty < \epsilon \quad (12)$$

**Definition 17 (Lipschitz condition, constant, and contraction).** Let  $D$  be a closed subset of  $\mathbb{R}^n$  and  $g : D \rightarrow D$ . If there exists a positive constant  $L$  such that

$$\|g(x) - g(y)\|_\infty \leq L\|x - y\|_\infty \quad (13)$$

for all  $x, y \in D$ , then  $g$  satisfies the Lipschitz condition on  $D$  in the  $\infty$ -norm.  $L$  is called the Lipschitz constant. If  $L \in (0, 1)$ , then  $g$  is called a contraction on  $D$  in the  $\infty$ -norm.

**Definition 18 (Jacobian).** Let  $g = (g_1, \dots, g_n)^T : \mathbb{R}^n \rightarrow \mathbb{R}^n$  be a function defined and continuous in an (open) neighborhood of  $\boldsymbol{\xi} \in \mathbb{R}^n$ . Suppose the first partial derivatives of each  $g_i$  exist at  $\boldsymbol{\xi}$ . The Jacobian matrix  $J_g(\boldsymbol{\xi})$  of  $g$  at  $\boldsymbol{\xi}$  is the  $n \times n$  matrix with elements

$$J_g(\boldsymbol{\xi})_{ij} = \frac{\partial g_i}{\partial x_j}(\boldsymbol{\xi}) \quad (14)$$

**Definition 19 (Newton's Method).** The sequence defined by

$$\mathbf{x}^{(k+1)} = \mathbf{x}^{(k)} - [J_f(\mathbf{x}^{(k)})]^{-1} f(\mathbf{x}^{(k)}) \quad (15)$$

where  $\mathbf{x}^{(0)} \in \mathbb{R}^n$ , is called Newton's method.

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**Algorithm 2** Power Iteration

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**Require:**  $v^{(0)}$  = some vector with  $\|v^{(0)}\| = 1$

- 1: **for**  $k \leftarrow 1, 2, \dots$  **do**
  - 2:      $w \leftarrow Av^{(k-1)}$  ▷ Apply  $A$
  - 3:      $v^{(k)} \leftarrow w/\|w\|$  ▷ Normalize
  - 4:      $\lambda^{(k)} \leftarrow (v^{(k)})^T Av^{(k)} = \langle v^{(k)}, Av^{(k)} \rangle$  ▷ Rayleigh Quotient
  - 5: **end for**
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**Algorithm 3** Simultaneous Iteration

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**Require:**  $Q^{(0)} = V = I$ , a list of vectors  $V$ , which we choose to be the identity

- 1: **for**  $k \leftarrow 1, 2, \dots$  **do**
  - 2:      $Z \leftarrow AQ^{(k-1)}$  ▷ Apply  $A$
  - 3:      $Z \leftarrow \underline{Q}^{(k)} R^{(k)}$  ▷  $QR$  factorization of  $Z$
  - 4:      $A^{(k)} \leftarrow (\underline{Q}^{(k)})^T A \underline{Q}^{(k)}$  ▷  $A_{ii}^{(k)} = \langle q_i^{(k)}, Aq_i^{(k)} \rangle$
  - 5: **end for**
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**Algorithm 4** QR Algorithm (without shifts)

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**Require:**  $A^{(0)} = A$

- 1: **for**  $k \leftarrow 1, 2, \dots$  **do**
  - 2:      $Q^{(k)} R^{(k)} \leftarrow A^{(k-1)}$  ▷  $QR$  factorization of  $A^{(k-1)}$
  - 3:      $A^{(k)} \leftarrow R^{(k)} Q^{(k)}$  ▷ Recombine factors in reverse order
  - 4: **end for**
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**Definition 20 (Lagrange basis polynomial).** Given the data  $\{x_i\}_{i=0}^n$ , define

$$l_j(x) = \frac{\prod_{i \neq j} (x - x_i)}{\prod_{i \neq j} (x_j - x_i)} \quad (16)$$

which satisfies

$$l_j(x_i) = \delta_{ij} = \begin{cases} 1 & i = j \\ 0 & i \neq j \end{cases} \quad (17)$$

(note that  $\prod_{i \neq j} (x - x_i)$  is an  $n$ th order polynomial (1 less degree than the number of data points) and  $\prod_{i \neq j} (x_j - x_i)$  is a constant).

**Definition 21 (Lagrange interpolation polynomial).** Given the data  $\{x_i\}_{i=0}^n$  and corresponding function values  $\{f(x_i)\}_{i=0}^n$  the Lagrange interpolation polynomial is

$$p(x) = \sum_{i=0}^n f(x_i) l_i(x) \quad (18)$$

**Definition 22 (Orthogonal polynomials).** Given a domain  $[a, b]$  and a weight function  $w(x)$  on the domain, a set of orthogonal polynomials is a list of polynomials  $\phi_0, \phi_1, \dots, \phi_N, \dots$  such that

$$\langle \phi_i, \phi_j \rangle = \int_a^b \phi_i(x) \phi_j(x) w(x) dx = \delta_{ij} \quad (19)$$

**Definition 23 (Autonomous).** If the force  $\mathbf{f}$  has no explicit dependence on  $t$ , then we call the ODE (system) autonomous.

**Definition 24 (Lipshitz continuous).** If

$$|f(u) - f(u^*)| \leq L|u - u^*| \quad (20)$$

for  $u$  in a small neighborhood of  $u^*$ , then  $f$  is Lipshitz continuous at  $u^*$ . Note that if  $f'$  exists, then

$$L = |f'(u^*)| \quad (21)$$

**Definition 25 (Uniformly Lipshitz continuous).** If  $L_u$  has an upper bound in the domain of  $f$ , then  $f$  is uniformly Lipshitz continuous.

**Definition 26 (Local Truncation Error (LTE)).** The local truncation error is by how much the true solution fails to satisfy the approximation scheme, which can be written as

$$\tau_n = \frac{u_{n+1} - u_n}{\Delta t} - f(u_n) \quad (22)$$

**Definition 27 (Consistency).** We say a method is consistent is the LTE goes to 0 as  $\Delta \rightarrow 0$ .