

**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

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1:  $L_1 \leftarrow \sqrt{a_{11}}$ 
2: for  $k \leftarrow 2, 3, \dots, n$  do
3:   Solve  $L_{k-1}l_k = a_k$  for  $l_k$ 
4:    $l_{kk} \leftarrow \sqrt{a_{kk} - l_k^T l_k}$ 
5:    $L_k \leftarrow \begin{pmatrix} L_{k-1} & 0 \\ l_k^T & l_{kk} \end{pmatrix}$ 
6: 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:      $A^{(k)} \leftarrow (\underline{Q}^{(k)})^T A \underline{Q}^{(k)}$
  - 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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