Fundamentals

- Basic Operations:
 - ♦ Matrix-vector product b = Ax, where $A \in \mathbb{R}^{m \times n}$ and $x \in \mathbb{R}^n$, is defined by

$$b_i = \sum_{j=1}^n a_{ij} x_j, \ i = 1, \dots, m$$

♦ Matrix-matrix product B = AC, where $A \in R^{\ell \times m}$ and $C \in R^{m \times n}$, is defined by

$$b_{ij} = \sum_{k=1}^{m} a_{ik} c_{kj}, \ i = 1, \dots, \ell, \ j = 1, \dots, n.$$

♦ Outer product: Let the columns of A be denoted as $A = [A_1, ..., A_n]$. Then b = Ax can be written as

$$b = \sum_{j=1}^{n} A_j x_j,$$

i.e., b is a linear combination of columns of A.

- Basic Definitions: Let $A \in \mathbb{R}^{m \times n}$. Then
 - $\diamond \ R(A) = \text{range space of } A = \{Ax \in R^m | x \in R^n\}.$
 - $\diamond \ N(A) = \text{null space of } A = \{ x \in R^n | Ax = 0 \}.$
 - \triangleright Note that both R(A) and N(A) are linear subspaces. Thus they have bases and dimensionalities.
 - \diamond rank(A) = the dimension of R(A).
 - \triangleright Matrix A has full rank if and only if $m \ge n$ and A is a one-to-one map.

- Inverse and Nonsingularity:
 - \diamond The operation Ax can be considered as a linear map from \mathbb{R}^n to \mathbb{R}^m .
 - ◊ Suppose A is a square matrix. The following results are equivalent:
 ▷ A has an inverse A⁻¹.
 - \triangleright rank(A) = n.
 - $\triangleright R(A) = R^n.$
 - $\triangleright N(A) = \{0\}.$
 - $\triangleright \det(A) \neq 0.$
 - $\triangleright Ax = b$ has a unique solution for every b.
 - \triangleright All *n* columns of *A* are linearly independent.
- Orthogonal Vectors:
 - ♦ We say two vectors $x, y \in \mathbb{R}^n$ are *orthogonal* to each other if and only if $x^T y = 0$.
 - ◇ The vectors in an orthogonal set are linearly independent.
 - ◇ A square matrix $A ∈ R^{n × n}$ is called an orthogonal matrix if all its columns are *orthonormal*, i.e., if $A_i^T A_i = \delta_{ij}$.
 - \triangleright A is orthogonal if and only if $A^T A = I$, i.e., $A^{-1} = A^T$.
- Norms:
 - \diamond The essential notions of size and distance in a vector space are captured by *norms*.
 - ♦ Suppose $x \in \mathbb{R}^n$. Then following are some commonly used vector norms for x.

$$\begin{aligned} ||x||_{1} &:= \sum_{i=1}^{n} |x_{i}|, \\ ||x||_{2} &:= \left(\sum_{i=1}^{n} |x_{i}|^{2}\right)^{1/2}, \\ ||x||_{\infty} &:= \max_{1 \le i \le n} |x_{i}|, \\ ||x||_{p} &:= \left(\sum_{i=1}^{n} |x_{i}|^{p}\right)^{1/p}, \ (1 \le p < \infty). \end{aligned}$$

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