

A: Basic vector and matrix operations

A.1 Vector operations

Vector addition: The addition of two N -dimensional vectors

$$\mathbf{a}_{N \times 1} = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_N \end{bmatrix} \quad \text{and} \quad \mathbf{b}_{N \times 1} = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_N \end{bmatrix}, \quad (\text{A.1})$$

is defined as the entry-wise sum of \mathbf{a} and \mathbf{b} , resulting in a vector of the same dimension denoted by

$$\mathbf{a} + \mathbf{b} = \begin{bmatrix} a_1 + b_1 \\ a_2 + b_2 \\ \vdots \\ a_N + b_N \end{bmatrix}. \quad (\text{A.2})$$

Subtraction of two vectors is defined in a similar fashion.

Vector multiplication by a scalar: Multiplying a vector \mathbf{a} by a scalar γ returns a vector of the same dimension whose every element is multiplied by γ

$$\gamma \mathbf{a} = \begin{bmatrix} \gamma a_1 \\ \gamma a_2 \\ \vdots \\ \gamma a_N \end{bmatrix}. \quad (\text{A.3})$$

Vector transpose: The transpose of a *column vector* \mathbf{a} (with vertically stored elements) is a *row vector* with the same elements which are now stored horizontally, denoted by

$$\mathbf{a}^T = [a_1 \quad a_2 \quad \cdots \quad a_N]. \quad (\text{A.4})$$

Similarly the transpose of a row vector is a column vector, and we have in general that $(\mathbf{a}^T)^T = \mathbf{a}$.

Inner product of two vectors: The inner product (or dot product) of two vectors \mathbf{a} and \mathbf{b} (of the same dimensions) is simply the sum of their component-wise product, written as

$$\mathbf{a}^T \mathbf{b} = \sum_{n=1}^N a_n b_n. \quad (\text{A.5})$$

The inner product of \mathbf{a} and \mathbf{b} is also often written as $\langle \mathbf{a}, \mathbf{b} \rangle$.

Inner product rule and correlation: The inner product rule between two vectors provides a measurement of

$$\mathbf{a}^T \mathbf{b} = \|\mathbf{a}\|_2 \|\mathbf{b}\|_2 \cos(\theta), \quad (\text{A.6})$$

where θ is the angle between the vectors \mathbf{a} and \mathbf{b} . Dividing both vectors by their length gives the *correlation* between the two vectors

$$\frac{\mathbf{a}^T \mathbf{b}}{\|\mathbf{a}\|_2 \|\mathbf{b}\|_2} = \cos(\theta), \quad (\text{A.7})$$

which ranges between -1 and 1 (when the vectors point in completely opposite or parallel directions respectively).

Outer product of two vectors: The outer product of two vectors $\mathbf{a}_{N \times 1}$ and $\mathbf{b}_{M \times 1}$ is an $N \times M$ matrix \mathbf{C} defined as

$$\mathbf{C} = \begin{bmatrix} a_1 b_1 & a_1 b_2 & \cdots & a_1 b_M \\ a_2 b_1 & a_2 b_2 & \cdots & a_2 b_M \\ \vdots & \vdots & \ddots & \vdots \\ a_N b_1 & a_N b_2 & \cdots & a_N b_M \end{bmatrix}. \quad (\text{A.8})$$

The outer product of \mathbf{a} and \mathbf{b} is also written as $\mathbf{a}\mathbf{b}^T$. Unlike the inner product, the outer product does not hold the commutative property meaning that the outer product of \mathbf{a} and \mathbf{b} does *not* necessarily equal the outer product of \mathbf{b} and \mathbf{a} .

A.2 Matrix operations

Matrix addition: The addition of two $N \times M$ matrices

$$\mathbf{A}_{N \times M} = \begin{bmatrix} a_{1,1} & a_{1,2} & \cdots & a_{1,M} \\ a_{2,1} & a_{2,2} & \cdots & a_{2,M} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N,1} & a_{N,2} & \cdots & a_{N,M} \end{bmatrix} \quad \text{and} \quad \mathbf{B}_{N \times M} = \begin{bmatrix} b_{1,1} & b_{1,2} & \cdots & b_{1,M} \\ b_{2,1} & b_{2,2} & \cdots & b_{2,M} \\ \vdots & \vdots & \ddots & \vdots \\ b_{N,1} & b_{N,2} & \cdots & b_{N,M} \end{bmatrix}, \quad (\text{A.9})$$

is defined again as the entry-wise sum of \mathbf{A} and \mathbf{B} , given as

$$\mathbf{A} + \mathbf{B} = \begin{bmatrix} a_{1,1} + b_{1,1} & a_{1,2} + b_{1,2} & \cdots & a_{1,M} + b_{1,M} \\ a_{2,1} + b_{2,1} & a_{2,2} + b_{2,2} & \cdots & a_{2,M} + b_{2,M} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N,1} + b_{N,1} & a_{N,2} + b_{N,2} & \cdots & a_{N,M} + b_{N,M} \end{bmatrix}. \quad (\text{A.10})$$

Subtraction of two matrices is defined in a similar fashion.

Matrix multiplication by a scalar: Multiplying a matrix \mathbf{A} by a scalar γ returns a matrix of the same dimension whose every element is multiplied by γ

$$\gamma \mathbf{A} = \begin{bmatrix} \gamma a_{1,1} & \gamma a_{1,2} & \cdots & \gamma a_{1,M} \\ \gamma a_{2,1} & \gamma a_{2,2} & \cdots & \gamma a_{2,M} \\ \vdots & \vdots & \ddots & \vdots \\ \gamma a_{N,1} & \gamma a_{N,2} & \cdots & \gamma a_{N,M} \end{bmatrix}. \quad (\text{A.11})$$

Matrix transpose: The transpose of an $N \times M$ matrix \mathbf{A} is formed by putting the transpose of each column of \mathbf{A} into the corresponding row of \mathbf{A}^T , giving the $M \times N$ transpose matrix as

$$\mathbf{A}^T = \begin{bmatrix} a_{1,1} & a_{2,1} & \cdots & a_{N,1} \\ a_{1,2} & a_{2,2} & \cdots & a_{N,2} \\ \vdots & \vdots & \ddots & \vdots \\ a_{1,M} & a_{2,M} & \cdots & a_{N,M} \end{bmatrix}. \quad (\text{A.12})$$

Again, we have $(\mathbf{A}^T)^T = \mathbf{A}$.

Matrix multiplication: The product of two matrices $\mathbf{A}_{N \times M}$ and $\mathbf{B}_{M \times P}$ is an $N \times P$ matrix defined via the sum of M outer product matrices, as

$$\mathbf{C} = \mathbf{AB} = \sum_{m=1}^M \mathbf{a}_m \mathbf{b}^m, \quad (\text{A.13})$$

where \mathbf{a}_m and \mathbf{b}^m respectively denote the m th column of \mathbf{A} and the m th row of \mathbf{B} .

The p th column of \mathbf{C} can be found via multiplying \mathbf{A} by the p th column of \mathbf{B} ,

$$\mathbf{c}_p = \mathbf{A} \mathbf{b}_p = \sum_{m=1}^M \mathbf{a}_m b_{m,p}. \quad (\text{A.14})$$

The n th row of \mathbf{C} can be found via multiplying the n th row of \mathbf{A} by \mathbf{B} ,

$$\mathbf{c}^n = \mathbf{a}^n \mathbf{B} = \sum_{m=1}^M a_{n,m} \mathbf{b}^m. \quad (\text{A.15})$$

The (n, p) th entry of \mathbf{C} is found by multiplying the n th row of \mathbf{A} by the p th column of \mathbf{B} ,

$$c_{n,p} = \mathbf{a}^n \mathbf{b}_p. \quad (\text{A.16})$$

Note that vector inner and outer products are special cases of matrix multiplication.

Entry-wise product: The entry-wise product (or Hadamard product) of two matrices $\mathbf{A}_{N \times M}$ and $\mathbf{B}_{N \times M}$ is defined as

$$\mathbf{A} \circ \mathbf{B} = \begin{bmatrix} a_{1,1}b_{1,1} & a_{1,2}b_{1,2} & \cdots & a_{1,M}b_{1,M} \\ a_{2,1}b_{2,1} & a_{2,2}b_{2,2} & \cdots & a_{2,M}b_{2,M} \\ \vdots & \vdots & \ddots & \vdots \\ a_{N,1}b_{N,1} & a_{N,2}b_{N,2} & \cdots & a_{N,M}b_{N,M} \end{bmatrix}. \quad (\text{A.17})$$

In other words, the (n, m) th entry of $\mathbf{A} \circ \mathbf{B}$ is simply the product of the (n, m) th entry of \mathbf{A} and the (n, m) th entry of \mathbf{B} .