

Basic Expansions for Linear Algebra Expressions

The notes below contain simple ways to rewrite matrix-vector and matrix-matrix products that we will use repeatedly throughout the course.

Basic Notation

For an $M \times N$ matrix \mathbf{A} , we denote the columns as $\mathbf{a}_{c1}, \dots, \mathbf{a}_{cN} \in \mathbb{R}^M$ and the rows as $\mathbf{a}_{r1}, \dots, \mathbf{a}_{rM} \in \mathbb{R}^N$, and so

$$\mathbf{A} = \begin{bmatrix} | & | & \cdots & | \\ \mathbf{a}_{c1} & \mathbf{a}_{c2} & \cdots & \mathbf{a}_{cN} \\ | & | & & | \end{bmatrix} = \begin{bmatrix} - & \mathbf{a}_{r1}^T & - \\ - & \mathbf{a}_{r2}^T & - \\ & \vdots & \\ - & \mathbf{a}_{rM}^T & - \end{bmatrix}.$$

We will often refer to the $\{\mathbf{a}_{rm}\}$ as “the rows of \mathbf{A} ”, even though, strictly speaking, the $\{\mathbf{a}_{rm}\}$ are column vectors in \mathbb{R}^N , and it is the $\{\mathbf{a}_{rm}^T\}$ that are the $1 \times N$ rows of \mathbf{A} .

The entries of a vector in \mathbb{R}^N and an $M \times N$ matrix will be denoted using brackets:

$$\mathbf{x} = \begin{bmatrix} x[1] \\ x[2] \\ \vdots \\ x[N] \end{bmatrix}, \quad \mathbf{A} = \begin{bmatrix} A[1,1] & A[1,2] & \cdots & A[1,N] \\ A[2,1] & A[2,2] & \cdots & A[2,N] \\ \vdots & & \ddots & \\ A[M,1] & A[M,2] & \cdots & A[M,N] \end{bmatrix}.$$

Notes that vectors \mathbf{x} and matrices \mathbf{A} are typeset in bold, while their entries $x[n]$ and $A[m,n]$ are not, since they are scalars.

Matrix-vector multiplies

We can think of the action of an $M \times N$ matrix \mathbf{A} on a vector $\mathbf{x} \in \mathbb{R}^N$ in one of two ways.

The first is as a series of inner products against the rows of \mathbf{A} :

$$\mathbf{Ax} = \begin{bmatrix} \mathbf{a}_{r1}^T \mathbf{x} \\ \mathbf{a}_{r2}^T \mathbf{x} \\ \vdots \\ \mathbf{a}_{rM}^T \mathbf{x} \end{bmatrix}.$$

The other is as a linear combination of the columns of \mathbf{A} :

$$\mathbf{Ax} = \sum_{n=1}^N x[n] \mathbf{a}_{cn}.$$

Matrix-matrix multiplies

Likewise, the product of an $M \times N$ matrix \mathbf{A} and a $N \times P$ matrix \mathbf{B} can be thought of as a collection of the inner products between all of the rows of \mathbf{A} and all of the columns of \mathbf{B} ,

$$\mathbf{AB} = \begin{bmatrix} \mathbf{a}_{r1}^T \mathbf{b}_{c1} & \mathbf{a}_{r1}^T \mathbf{b}_{c2} & \cdots & \mathbf{a}_{r1}^T \mathbf{b}_{cP} \\ \mathbf{a}_{r2}^T \mathbf{b}_{c1} & \mathbf{a}_{r2}^T \mathbf{b}_{c2} & \cdots & \mathbf{a}_{r2}^T \mathbf{b}_{cP} \\ \vdots & & \ddots & \\ \mathbf{a}_{rM}^T \mathbf{b}_{c1} & \mathbf{a}_{rM}^T \mathbf{b}_{c2} & \cdots & \mathbf{a}_{rM}^T \mathbf{b}_{cP} \end{bmatrix},$$

as a sum of the rank 1 matrices formed by taking the outer product of the columns of \mathbf{A} with the rows of \mathbf{B} ,

$$\mathbf{AB} = \sum_{n=1}^N \mathbf{a}_{cn} \mathbf{b}_{rn}^T,$$

as left action of \mathbf{A} on the collective columns of \mathbf{B} ,

$$\mathbf{AB} = \begin{bmatrix} \left| \begin{array}{c} \mathbf{A} \mathbf{b}_{c1} \\ \mathbf{A} \mathbf{b}_{c2} \\ \vdots \\ \mathbf{A} \mathbf{b}_{cP} \end{array} \right| \end{bmatrix}$$

or as right action of \mathbf{B} on the rows of \mathbf{A}

$$\mathbf{AB} = \begin{bmatrix} - & \mathbf{a}_{r1}^T \mathbf{B} & - \\ - & \mathbf{a}_{r2}^T \mathbf{B} & - \\ & \vdots & \\ - & \mathbf{a}_{rM}^T \mathbf{B} & - \end{bmatrix}.$$

We again stress that these are just four different ways to write down exactly the same thing.

Second-order forms

For an $N \times N$ matrix \mathbf{A} and a vector $\mathbf{x} \in \mathbb{R}^N$, the quadratic form $\mathbf{x}^T \mathbf{A} \mathbf{x}$ can be expanded as

$$\mathbf{x}^T \mathbf{A} \mathbf{x} = \sum_{m=1}^N \sum_{n=1}^N A[m, n] x[m] x[n].$$

Similarly, for an $M \times N$ matrix \mathbf{A} and vectors $\mathbf{y} \in \mathbb{R}^M$, $\mathbf{x} \in \mathbb{R}^N$, the bilinear form $\mathbf{y}^T \mathbf{A} \mathbf{x}$ can be expanded as

$$\mathbf{y}^T \mathbf{A} \mathbf{x} = \sum_{m=1}^M \sum_{n=1}^N A[m, n] y[m] x[n].$$

Note that if \mathbf{D} is an $N \times N$ diagonal matrix, so $D[m, n] = 0$ for $m \neq n$, then

$$\mathbf{x}^T \mathbf{D} \mathbf{x} = \sum_{n=1}^N D[n, n] x[n]^2.$$

Three matrices

Let \mathbf{U} be an $M \times N$ matrix, \mathbf{C} a $N \times P$ matrix, and \mathbf{W} a $P \times Q$ matrix. Then the $M \times Q$ matrix \mathbf{UCW} can be written as

$$\mathbf{UCW} = \begin{bmatrix} \mathbf{u}_{r1}^T \mathbf{C} \mathbf{w}_{c1} & \mathbf{u}_{r1}^T \mathbf{C} \mathbf{w}_{c2} & \cdots & \mathbf{u}_{r1}^T \mathbf{C} \mathbf{w}_{cQ} \\ \mathbf{u}_{r2}^T \mathbf{C} \mathbf{w}_{c1} & \mathbf{u}_{r2}^T \mathbf{C} \mathbf{w}_{c2} & \cdots & \mathbf{u}_{r2}^T \mathbf{C} \mathbf{w}_{cQ} \\ \vdots & & \ddots & \\ \mathbf{u}_{rM}^T \mathbf{C} \mathbf{w}_{c1} & \mathbf{u}_{rM}^T \mathbf{C} \mathbf{w}_{c2} & \cdots & \mathbf{u}_{rM}^T \mathbf{C} \mathbf{w}_{cQ} \end{bmatrix},$$

or

$$\mathbf{UCW} = \sum_{n=1}^N \sum_{p=1}^P C[n, p] \mathbf{u}_{cn} \mathbf{w}_{rp}^T.$$

In the special case where \mathbf{C} is square and diagonal

$$\mathbf{C} = \begin{bmatrix} c_1 & & & \\ & c_2 & & \\ & & \ddots & \\ & & & c_N \end{bmatrix},$$

then the above reduces to

$$\mathbf{UCW} = \sum_{n=1}^N c_n \mathbf{u}_{cn} \mathbf{w}_{rn}^T.$$