

July 15 2019.

3.C.

Addition and scalar multiplication of matrices.

3.35 Definition.

Matrix addition is the sum of two matrices of the same size, obtained adding the corresponding entries in the matrices.

$$\begin{bmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & \ddots & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{bmatrix} + \begin{bmatrix} C_{1,1} & \cdots & C_{1,n} \\ \vdots & \ddots & \vdots \\ C_{m,1} & \cdots & C_{m,n} \end{bmatrix} = \begin{bmatrix} A_{1,1}+C_{1,1} & \cdots & A_{1,n}+C_{1,n} \\ \vdots & \ddots & \vdots \\ A_{m,1}+C_{m,1} & \cdots & A_{m,n}+C_{m,n} \end{bmatrix}$$

In other words, $A_{j,k} + C_{j,k} = (A+C)_{j,k}$ for each $j=1, \dots, m$, $k=1, \dots, n$

3.36 The matrix of the sum of linear maps.

Suppose $S, T \in \mathcal{L}(V, W)$. Then $M(S+T) = M(S) + M(T)$

3.37 Definition

Scalar multiplication of a matrix is the product of a scalar and a matrix, obtained by multiplying each entry in the matrix by the scalar:

$$\lambda \begin{bmatrix} A_{1,1} & \cdots & A_{1,n} \\ \vdots & \ddots & \vdots \\ A_{m,1} & \cdots & A_{m,n} \end{bmatrix} = \begin{bmatrix} \lambda A_{1,1} & \cdots & \lambda A_{1,n} \\ \vdots & \ddots & \vdots \\ \lambda A_{m,1} & \cdots & \lambda A_{m,n} \end{bmatrix}$$

In other words, $(\lambda A)_{j,k} = \lambda A_{j,k}$ for each $j=1, \dots, m$; $k=1, \dots, n$.

3.38 The matrix of a scalar times a linear map.

Suppose $\lambda \in F$ and $T \in \mathcal{L}(V, W)$. Then $M(\lambda T) = \lambda M(T)$

3.40 $\dim F^{m,n} = mn$

Let m and n be positive integers. Let $F^{m,n}$ is the set of all $m \times n$ matrices with entries in F . Then $\dim F^{m,n} = mn$.

Proof:

$F^{m,n}$ is a vector space with respect to operations of matrix addition and scalar multiplication of a matrix.

Now $\dim F^{m,n} = mn$ because

$$\begin{aligned}
 & \left[\begin{smallmatrix} 1 & \cdots & 0 \\ \vdots & & \vdots \\ 0 & \cdots & 0 \end{smallmatrix} \right], \left[\begin{smallmatrix} 0 & 1 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & \cdots & 0 & 1 \end{smallmatrix} \right], \dots, \left[\begin{smallmatrix} 0 & \cdots & 1 \\ \vdots & & \vdots \\ 0 & \cdots & 0 \end{smallmatrix} \right], & \begin{array}{l} m\text{-rows} \\ n\text{-columns} \end{array} \\
 & \underbrace{\hspace{10em}}_{n \text{ matrices}} \\
 & \left[\begin{smallmatrix} 0 & \cdots & 0 \\ 1 & 0 & \cdots & 0 \\ 0 & \cdots & 0 \end{smallmatrix} \right], \left[\begin{smallmatrix} 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & \cdots & 0 & 1 \end{smallmatrix} \right], \dots, \left[\begin{smallmatrix} 0 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & \cdots & 0 & 1 \end{smallmatrix} \right] \\
 & \vdots \\
 & \left[\begin{smallmatrix} 0 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 1 & \cdots & 0 \end{smallmatrix} \right], \left[\begin{smallmatrix} 0 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & 1 & \cdots & 0 \end{smallmatrix} \right], \dots, \left[\begin{smallmatrix} 0 & \cdots & 0 \\ \vdots & & \ddots & \vdots \\ 0 & \cdots & 0 & 1 \end{smallmatrix} \right]
 \end{aligned}$$

 is a basis of $F^{m,n}$ so $\dim F^{m,n} = m \cdot n$.

Matrix multiplication

Let $v_1, \dots, v_n$ be a basis of V and $w_1, \dots, w_m$ be a basis of W .

Also let U be a vector space and $u_1, \dots, u_p$ is a basis of U .

Let $T: U \rightarrow V$ and $S: V \rightarrow W$ be linear maps.

We will show: $M(ST) = M(S)M(T)$.

Suppose $M(S) = A$ and $M(T) = C$. For each $k=1, \dots, p$ we have

$$(ST)u_k = S\left(\sum_{r=1}^n C_{r,k} v_r\right) \quad \text{by Definition 3.32 Axler}$$

$$= \sum_{r=1}^n C_{r,k} v_r \quad \text{Since } S \text{ is linear.}$$

$$= \sum_{r=1}^n C_{r,k} \left(\sum_{j=1}^m A_{j,r} w_j \right)$$

$$= \sum_{j=1}^m \left(\sum_{r=1}^n A_{j,r} C_{r,k} \right) w_j$$

By definition 3.32 of Axler, $M(ST)$ is the $m \times p$ matrix with entry in row j , column k .

$$D_{j,k} = \sum_{r=1}^n A_{j,r} C_{r,k}.$$

3.41 Definition

Let A be an $m \times n$ matrix and C be an $n \times p$ matrix.

Then AC is an $m \times p$ matrix with entry row j , column k .

$$(AC)_{j,k} = \sum_{r=1}^n A_{j,r} C_{r,k}$$

In other words,

$$\begin{aligned} AC &= \begin{bmatrix} A_{1,1} & \dots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \dots & A_{m,n} \end{bmatrix} \cdot \begin{bmatrix} C_{1,1} & \dots & C_{1,p} \\ \vdots & & \vdots \\ C_{n,1} & \dots & C_{n,p} \end{bmatrix} \\ &= \begin{bmatrix} \sum_{r=1}^n A_{1,r} C_{r,1} & \dots & \sum_{r=1}^n A_{1,r} C_{r,p} \\ \vdots & & \vdots \\ \sum_{r=1}^n A_{m,r} C_{r,1} & \dots & \sum_{r=1}^n A_{m,r} C_{r,p} \end{bmatrix} \end{aligned}$$

3.42 Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} \cdot \begin{bmatrix} 6 & 5 & 4 & 3 \\ 2 & 1 & 0 & -1 \end{bmatrix} = \begin{bmatrix} 10 & 7 & 4 & 1 \\ 26 & 19 & 12 & 5 \end{bmatrix}$$

$$\begin{bmatrix} 5 & 6 \end{bmatrix} \quad 2 \times 4 \quad \begin{bmatrix} 42 & 31 & 20 & 9 \end{bmatrix}$$

$3 \times 2 \qquad \qquad \qquad 3 \times 4$

3.43 The matrix of the product of linear maps.

If $T \in \mathcal{L}(U, V)$ and $S \in \mathcal{L}(V, W)$ then

$$M(ST) = M(S)M(T)$$

Let A be an $m \times n$ matrix.

$$A = \begin{bmatrix} A_{1,1} & \cdots & A_{1,k} & \cdots & A_{1,n} \\ \vdots & & \vdots & & \vdots \\ A_{j,1} & \cdots & A_{j,k} & \cdots & A_{j,n} \\ \vdots & & \vdots & & \vdots \\ A_{m,1} & \cdots & A_{m,k} & \cdots & A_{m,n} \end{bmatrix}$$

Then $A_{j\cdot} = (A_{j,1} \cdots A_{j,n})$

$$A_{\cdot,k} = \begin{bmatrix} A_{1,k} \\ \vdots \\ A_{m,k} \end{bmatrix}$$

3.45 Example

$$\text{Let } A = \begin{bmatrix} 8 & 4 & 5 \\ 1 & 9 & 7 \end{bmatrix} \quad \text{Then } A_{2\cdot} = [1 \ 9 \ 7]$$

$$A_{\cdot,2} = \begin{bmatrix} 4 \\ 9 \end{bmatrix}$$

3.46 Example

$$\text{Let } A = (3 \ 4) \text{ and } C = \begin{pmatrix} 6 \\ 2 \end{pmatrix}$$

$$AC = (3 \ 4) \begin{pmatrix} 6 \\ 2 \end{pmatrix} = (26) = 26$$

$$(AC)_{1\cdot} = 26 \quad , \quad (AC)_{\cdot,1} = 26 \quad , \quad (AC)_{\cdot,1} = 26$$

3.47 Entry of matrix product equals row times column.

Suppose A is an $m \times n$ matrix and C is an $n \times p$ matrix.

Then $(AC)_{j,k} = A_{j,\cdot} \cdot C_{\cdot,k}$

$$\begin{aligned} \text{Proof: } (AC)_{j,k} &= \left(\begin{bmatrix} A_{j,1} & \dots & A_{j,n} \\ A_{m,1} & \dots & A_{m,n} \end{bmatrix} \begin{bmatrix} C_{1,1} & \dots & C_{1,p} \\ \vdots & & \vdots \\ C_{n,1} & \dots & C_{n,p} \end{bmatrix} \right)_{j,k} \\ &= \begin{vmatrix} \sum_{r=1}^n A_{j,r} C_{r,1} & \dots & \sum_{r=1}^n A_{j,r} C_{r,p} \\ \vdots & & \vdots \\ \sum_{r=1}^n A_{m,r} C_{r,1} & \dots & \sum_{r=1}^n A_{m,r} C_{r,p} \end{vmatrix}_{j,k} \\ &= \sum_{r=1}^n A_{j,r} C_{r,k} \end{aligned}$$

$$\begin{aligned} A_{j,\cdot} \cdot C_{\cdot,k} &= (A_{j,1} \dots A_{j,n}) \begin{bmatrix} C_{1,k} \\ \vdots \\ C_{n,k} \end{bmatrix} \\ &= \sum_{r=1}^n A_{j,r} C_{r,k} \end{aligned}$$

$$\therefore (AC)_{j,k} = A_{j,\cdot} \cdot C_{\cdot,k}$$

3.50 Example

$$\begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} \cdot \begin{bmatrix} 5 \\ 1 \end{bmatrix} = \begin{bmatrix} 7 \\ 19 \\ 31 \end{bmatrix}$$

$$3.41 \quad \begin{bmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{bmatrix} \cdot \begin{pmatrix} 6 & 5 & 4 & 3 \\ 2 & 1 & 0 & -1 \end{pmatrix} : AC = \begin{bmatrix} 10 & 7 & 4 & 1 \\ 26 & 19 & 12 & 5 \\ 42 & 31 & 20 & 9 \end{bmatrix}$$

3.52 Linear combination of columns.

Suppose A is an $m \times n$ matrix and $C = \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix}$ is an $n \times 1$ matrix.

$$\text{Then } AC = C_1 A_{\cdot,1} + \dots + C_n A_{\cdot,n}.$$

Proof:

$$\begin{aligned}
 C_1 A_{:,1} + \dots + C_n A_{:,n} &= C_1 \begin{pmatrix} A_{1,1} \\ \vdots \\ A_{m,1} \end{pmatrix} + \dots + C_n \begin{pmatrix} A_{1,n} \\ \vdots \\ A_{m,n} \end{pmatrix} \\
 &= \begin{pmatrix} C_1 A_{1,1} \\ \vdots \\ C_1 A_{m,1} \end{pmatrix} + \dots + \begin{pmatrix} C_n A_{1,n} \\ \vdots \\ C_n A_{m,n} \end{pmatrix} \\
 &= \begin{pmatrix} C_1 A_{1,1} + \dots + C_n A_{1,n} \\ \vdots \\ C_1 A_{m,1} + \dots + C_n A_{m,n} \end{pmatrix} \\
 AC &= \begin{pmatrix} A_{1,1} & \dots & A_{1,n} \\ \vdots & & \vdots \\ A_{m,1} & \dots & A_{m,n} \end{pmatrix} \begin{pmatrix} C_1 \\ \vdots \\ C_n \end{pmatrix} \\
 &= \begin{pmatrix} A_{1,1}C_1 + \dots + A_{1,n}C_n \\ \vdots \\ A_{m,1}C_1 + \dots + A_{m,n}C_n \end{pmatrix} \\
 \therefore AC &= C_1 A_{:,1} + \dots + C_n A_{:,n}
 \end{aligned}$$

3.D Invertibility and Isomorphic Vector Spaces.

Invertible linear maps.

3.53 Definition

- A linear map $T \in \mathcal{L}(V, W)$ is called invertible if there exists a linear map $S \in \mathcal{L}(W, V)$ s.t. $ST = I_V$ and $TS = I_W$, where I_V and I_W are identity maps on V and W respectively.

identity maps:

$$I_V: V \rightarrow V$$

$$I_V(v) = v$$

$$I_W: W \rightarrow W$$

$$I_W(w) = w$$

V, W vector spaces

v, w vectors in V, W

$$T: V \rightarrow W$$

$$S: W \rightarrow V$$

$$\therefore ST: V \rightarrow V$$

$$I_V: V \rightarrow V$$

If T is invertible with inverse S , then $ST = I_V$ and $TS = I_W$.

$$TS = I_W$$

$$S: W \rightarrow V$$

$$T: V \rightarrow W$$

$$\therefore TS: W \rightarrow W$$

$$(I_W: W \rightarrow W)$$

3.59 Inverse is unique

An invertible linear map has a unique inverse.

Proof:

Suppose $T \in \mathcal{L}(V, W)$ is invertible, and let S and $\tilde{S}$ be inverse of T .

$$\text{Then } S = SI_W$$

because $\tilde{S}$ is an inverse of T

$$= ST(\tilde{S})$$

$$= (ST)\tilde{S}$$

because S is an inverse of T .

$$= I_V \tilde{S}$$

$$= \tilde{S}$$

$\therefore S = \tilde{S}$ which means the inverse of T is unique.

If T is invertible, then its inverse is denoted by T^{-1} .

If $T \in \mathcal{L}(V, W)$ is invertible, then $T^{-1} \in \mathcal{L}(W, V)$ is the unique element s.t.

$$T^{-1}T = I_V, \text{ and } TT^{-1} = I_W.$$