## Section 2.2. Matrix Multiplication

Matrix multiplication is a little more complicated than matrix addition or scalar multiplication. If $A$ is an $m \times n$ matrix, then $B$ is an $n \times k$ matrix, 9 the product $A B$ of $A$ and $B$ is the $m \times k$ matrix whose $(i, j)$ - entry is computed as follow:
Multiply each entry of row $i$ of $A$ by the corresponding entry of column $j$ of $B$, and add the result
This is called the dot product of row $i$ of $A$ and column $j$ of $B$.

Example 20 Compute the $(1,3)$ and $(2,4)$ - entries of $A B$ where:

$$
A=\left[\begin{array}{ccc}
3 & -1 & 2 \\
0 & 1 & 4
\end{array}\right] \quad B=\left[\begin{array}{cccc}
2 & 1 & 6 & 0 \\
0 & 2 & 3 & 4 \\
-1 & 0 & 5 & 8
\end{array}\right]
$$

Then compute $A B$.

## Solution

The $(1,3)$ - entry of $A B$ is the dot product of row 1 of $A$ and column 3 of $B$, computed by multiplying corresponding entries and adding the result:

$$
3.6+(-1) .3+2.5=18-3+10=25
$$

Similarly, the $(2,4)$ entry of $A B$ is the dot product of row 2 of $A$ and column 4 of $B$, computed by multiplying corresponding entries and adding the result:

$$
0.0+1.4+4.8=0+4+32=36
$$

Since $A$ is $2 \times 3$ and $B$ is $3 \times 4$, the product is a $2 \times 4$ matrix:

$$
\left[\begin{array}{ccc}
3 & -1 & 2 \\
0 & 1 & 4
\end{array}\right]\left[\begin{array}{cccc}
2 & 1 & 6 & 0 \\
0 & 2 & 3 & 4 \\
-1 & 0 & 5 & 8
\end{array}\right]=\left[\begin{array}{cccc}
4 & 1 & 25 & 12 \\
-4 & 2 & 23 & 36
\end{array}\right]
$$

Computing the $(i, j)$ - entry of $A B$ involves going across row $i$ of $A$ and down column $j$ of $B$, multiplying corresponding entries, and adding the results. This requires that the rows of $A$ and the columns of $B$ be the same length. The following rule is a useful way to remember when the product of $A$ and $B$ can be formed and what the size of the product matrix is.

## Rule

Suppose $A$ and $B$ have sizes $m \times n$ and $n^{\prime} \times p$, respectively:

$$
m \times n \times p
$$

The product $A B$ can be formed only when $n=n^{\prime}$; in this case, the product matrix $A B$ is of size $m \times p$. When this happens, the product $A B$ is defined.

Example 21 If $A=【 \begin{array}{llll}- & 2_{-}^{-}\end{array}$and $B=\left[\begin{array}{l}5 \\ 6 \\ 4\end{array}\right]$, Compute $A^{2}, A B, B A, B^{2}$ when they are defined.

## Solution

Here, $A$ is a $1 \times 3$ matrix and $B$ is a $3 \times 1$ matrix, so $A^{2}, B^{2}$ are not defined. The $A B$ and $B A$ are defined, these are $1 \times 1$ and $3 \times 3$ matrices, respectively:

$$
A B=【 \quad 3 \quad 2_{-}^{-}\left[\begin{array}{l}
5 \\
6 \\
4
\end{array}\right]=\boldsymbol{I} \cdot 5+3.6+2.4=\mathbf{\$}+18+8=\boldsymbol{=} 1_{-}^{-}
$$

$$
B A=\left[\begin{array}{l}
5 \\
6 \\
4
\end{array}\right] 【 \quad 32_{-}^{-}=\left[\begin{array}{lll}
5.1 & 5.3 & 5.2 \\
6.1 & 6.3 & 6.2 \\
4.1 & 4.3 & 4.2
\end{array}\right]=\left[\begin{array}{ccc}
5 & 15 & 10 \\
6 & 18 & 12 \\
4 & 12 & 8
\end{array}\right]
$$

Unlike numerical multiplication, matrix products $A B$ and $B A$ need not be equal.
The number 1 plays a neutral role in numerical multiplication in the sense that $1 . a=a$ and $a .1=a$ for all number $a$. An analogous role for matrix multiplication is played by square matrices of the following types:

$$
\left\lceil\left[\begin{array}{ll}
1 & 0 \\
0 & 1
\end{array}\right],\left[\begin{array}{lll}
1 & 0 & 0 \\
0 & 1 & 0 \\
0 & 0 & 1
\end{array}\right]\right. \text { and so on. }
$$

In general, an identity matrix $I$ is a square matrix with $l$ 's on the main diagonal and zeros elsewhere. If it is important to stress the size of an $n \times n$ identity matrix, denoted by $I_{n}$. Identity matrix ply a neutral role with respect to matrix multiplication in the sense that :

$$
A I=A \text { and } I B=B
$$

whenever the product are defined
More formally, give the definition of matrix multiplication as follow:
If $A=\prod_{i j}$ is $m \times n$ and $B$ is $n \times p$ the $i$ th row of $A$ and the $j$ th column of $B$ are, respectively,

$$
\boldsymbol{I}_{i 1} \quad a_{i 2} \quad \ldots \quad a_{i n_{-}}{ }^{-} \text {and }\left[\begin{array}{c}
b_{1 j} \\
b_{2 j} \\
\ldots \\
b_{n j}
\end{array}\right]
$$

Hence, the $(i, j)$-entry of the product matrix $A B$ is the dot product:

$$
a_{i 1} b_{1 j}+a_{i 2} b_{2 j}+\ldots+a_{i n} b_{n j}=\sum_{k=1}^{n} a_{i k} b_{k j}
$$

This is useful in verifying fact about matrix multiplication

## Theorem 4

Assume that $k$ is an arbitrary scalar and that $A, B$ and $C$ are matrices of sizes such that the indicated operations can be performed:

1. $I A=A, B I=B$
2. $A(B C)=(A B) C$
3. $A(B+C)=A B+A C ; A(B-C)=A B-A C$
4. $(B+C) A=B A+C A ;(B-C) A=B A-C A$
5. $k(A B)=(k A) B=A(k B)$
6. $(A B)^{T}=B^{T} A^{T}$

## Matrices and Linear Equations

One of the most important motivations for matrix multiplication results from its close connection with systems of linear equations.
Consider any system of linear equations:

$$
\begin{aligned}
& a_{11} x_{1}+a_{12} x_{2}+\ldots+a_{l n} x_{n}=b_{1} \\
& a_{21} x_{1}+a_{22} x_{2}+\ldots+a_{2 n} x_{n}=b_{2} \\
& \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
& \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
& \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \ldots \\
& a_{m 1} x_{1}+a_{m 2} x_{2}+\ldots+a_{m n} x_{n}=b_{m}
\end{aligned}
$$

If $A=\left[\begin{array}{cccc}a_{11} & a_{12} & \ldots & a_{1 n} \\ a_{21} & a_{22} & \ldots & a_{2 n} \\ \ldots & \ldots & \ldots & \ldots \\ a_{m 1} & a_{m 2} & \ldots & a_{m n}\end{array}\right], \quad X=\left[\begin{array}{c}x_{1} \\ x_{2} \\ \ldots \\ x_{n}\end{array}\right], \quad B=\left[\begin{array}{c}b_{1} \\ b_{2} \\ \ldots \\ b_{m}\end{array}\right]$

These equations become the single matrix equation:

$$
A X=B
$$

This is called the matrix form of the system of equations, and $B$ is called the constant matrix. Matrix $A$ is called the coefficient matrix of the system of linear equations, and a column matrix $X_{l}$ is called a solution to the system if $A X_{I}=B$. The matrix form is useful for formulating results about solutions of system of linear equations. Given a system $A X=B$ there is a related system:

$$
A X=0
$$

called the associated homogenous system. If $X_{l}$ is a solution to $A X=B$ and if $X_{0}$ is a solution to $A X=0$, then $X_{1}+X_{0}$ is a solution to $A X=B$. Indeed, $A X_{1}=B$ and $A X_{0}=0$, so:

$$
A\left(X_{1}+X_{0}\right)=A X_{1}+A X_{0}=B+0=B
$$

This observation has a useful converse.

## Theorem 6

Suppose $X_{1}$ is a particular solution to a system $A X=B$ of linear equations.
Then every solution $X_{2}$ to $A X=B$ has the form :

$$
X_{2}=X_{0}+X_{1}
$$

for some solution $X_{0}$ of the associated homogeneous system $A X=0$.

## Proof:

Suppose that $X_{2}$ is any solution to $A X=B$ so that $A X_{2}=B$. Write $X_{0}=$ $X_{2}-X_{1}$, then $X_{2}=X_{0}+X_{1}$, and we compute:

$$
A X_{0}=A\left(X_{2}-X_{1}\right)=A X_{2}-A X_{1}=B-B=0
$$

Thus $X_{0}$ is a solution to the associated homogeneous system $A X=0$.

The important of Theorem 2 lies in the fact that sometimes a particular solution $X_{l}$ is easily to found, and so the problem of finding all solutions is reduced solving the associated homogeneous system.

Example 22 Express every solution to the following system as the sum of a specific solution plus a solution to the associated homogeneous system.

$$
\begin{aligned}
& x-y-z=2 \\
& 2 x-y-3 z=6 \\
& x-2 z=4
\end{aligned}
$$

## Solution

Gaussian elimination gives $x=4+2 t, y=2+t, z=t$, where $t$ is arbitrary. Hence the general solution is:

$$
X=\left[\begin{array}{l}
x \\
y \\
z
\end{array}\right]=\left[\begin{array}{c}
4+2 t \\
2+t \\
t
\end{array}\right]=\left[\begin{array}{l}
4 \\
2 \\
0
\end{array}\right]+\left[\begin{array}{l}
2 \\
1 \\
1
\end{array}\right] t
$$

Thus $X_{0}=\left[\begin{array}{l}4 \\ 2 \\ 0\end{array}\right]$ is a specific solution, and $X_{1}=\left[\begin{array}{l}2 \\ 1 \\ 1\end{array}\right] t$ gives all solutions to the associated homogeneous system (do the Gaussian elimination with all the constants zero)

Theorem 6 focuses attention on homogeneous systems. In that case there is a convenient matrix form for the solutions that will be needed later.

Example 23 Solve the homogeneous system $A X=0$, where:

$$
A=\left[\begin{array}{cccc}
1 & -2 & 3 & -2 \\
-3 & 6 & 1 & 0 \\
-2 & 4 & 4 & -2
\end{array}\right]
$$

## Solution

The reduction of the augmented matrix to reduced form is:

$$
\left[\begin{array}{cccc}
1 & -2 & 3 & -2 \\
-3 & 6 & 1 & 0 \\
-2 & 4 & 4 & -2
\end{array}\right] \rightarrow\left[\begin{array}{cccc}
1 & -2 & 0 & -\frac{1}{5} \\
0 & 0 & 1 & -\frac{3}{5} \\
0 & 0 & 0 & 0
\end{array}\right]
$$

So the solution are $a=2 r+\frac{1}{5} t, b=r, c=\frac{3}{5} t, d=t$ by Gaussian elimination. Hence we can write the general solution $X$ in the matrix form:

$$
X=\left[\begin{array}{l}
a \\
b \\
c \\
d
\end{array}\right]=\left[\begin{array}{c}
2 r+\frac{1}{5} t \\
r \\
\frac{3}{5} t \\
t
\end{array}\right]=r\left[\begin{array}{l}
2 \\
l \\
0 \\
0
\end{array}\right]+t\left[\begin{array}{c}
\frac{1}{5} \\
0 \\
\frac{3}{5} \\
l
\end{array}\right]=r X_{l}+t X_{2}
$$

Where $X_{1}=\mathbb{I} \quad 1 \quad 0 \quad 00_{-}^{-}$and $X_{2}=\begin{array}{lllll}0 & \frac{3}{5} & 1 \text {. are particular solutions }\end{array}$ determined by the Gaussian Algorithm.

The solutions $X_{1}$ and $X_{2}$ in Example 23 are called the basic solutions to the homogeneous system, and a solution of the form $r X_{1}+t X_{2}$ is called linear combination of the basic solution $X_{1}$ and $X_{2}$.

In the same way, the Gaussian algorithm produces basic solutions to every homogeneous system $A X=O$ (there are no basic solution if there is only the trivial solution ). Moreover, every solution is given by the algorithm as a linear combination of these basic solutions ( as in example 23 )

## Exercises 2.2

1. Find $a, b, c, d$ if :
a. $\left[\begin{array}{ll}a & b \\ c & d\end{array}\right]\left[\begin{array}{cc}3 & -5 \\ -1 & 2\end{array}\right]=\left[\begin{array}{cc}1 & -1 \\ 2 & 0\end{array}\right]$
b. $\left[\begin{array}{cc}2 & 1 \\ -1 & 2\end{array}\right]\left[\begin{array}{ll}a & b \\ c & d\end{array}\right]=\left[\begin{array}{cc}7 & 2 \\ -1 & 4\end{array}\right]$
2. Verify that $A^{2}-A-6 I=0$ if:
a. $A=\left[\begin{array}{ll}3 & -1 \\ 0 & -2\end{array}\right]$
b. $A=\left[\begin{array}{cc}2 & 2 \\ 2 & -1\end{array}\right]$
3. Express every solution of the system as a sum of a specific solution plus a solution of the associated homogenous system.
$x-y-4 z=-4$
a. $x+2 y+5 z=2$
$x+y+2 z=0$

$$
\text { b. } \begin{aligned}
& 2 a+b-c-d=-1 \\
& 3 a+b+c-2 d=-2 \\
& -a-b+2 c+d=2 \\
& -2 a-b+2 d=3
\end{aligned}
$$

4. Find the basic solutions and write the general solution as a linear combination of the basic solutions.
$a+2 b-c+2 d+e=0$
$a+b-2 c+3 d+2 e=0$
a. $a+2 b+2 c+e=0$
b. $2 a-b+3 c+4 d+e=0$
$2 a+4 b-2 c+3 d+e=0$
$-a-2 b+3 c+d=0$
$3 a+c+7 d+2 e=0$
5. Let $B$ be an $n \times n$ matrix. Suppose $A B=0$ for some non zero $m \times n$ matrix $A$. Show that no $n \times n$ matrix $C$ exists such that $B C=I$.
6. The trace of a square matrix $A$, denoted $\operatorname{tr} A$, is the sum of the elements on the main diagonal of $A$. Show that, if $A, B$ are $n \times n$ matrices:
a. $\operatorname{tr}(A+B)=\operatorname{tr} A+\operatorname{tr} B$
b. $\quad \operatorname{tr}(k A)=k t A$ for any number of $k$
C. $\operatorname{tr}\left(A^{T}\right)=\operatorname{tr} A$
d. $\quad \operatorname{tr}(A B)=\operatorname{tr}(B A)$
7. A square matrix is called idempotent if $P^{2}=P$. Show that:
a. $O, I$ are idempotents
b. If $P$ is idempotent, so is $I-P$ and $P(I-P)=0$
c. If $P$ is idempotent, so is $P^{T}$
8. If $P$ is an idempotent, so is $Q=P+A P-P A P$ for any square matrix $A$ ( of the same size as $P$ )
9. Let $A$ be $n \times m$ and $B$ be $m \times n$. If $A B=I$ then $B A$ is Idempotent.
10. Let $A$ and $B$ be $n \times n$ diagonal matrices (all entries off the main diagonal are zero ) ,show that $A B$ is diagonal and $A B=B A$
