

## *Parametric Vector Form of the Solution*

**ALGOR** The solution to a general linear system can be written in **parametric vector form** as: one vector plus an arbitrary linear combination of vectors satisfying the corresponding homogeneous system.

## Defintion

**Defn.** A collection of vectors is **linearly independent** if the only linear combination of them that equals  $0$  is the trivial combination (all weights zero). Otherwise it is said to be **linearly dependent**.

Note that the collection is linearly dependent if some vector in it can be written as a linear combination of the other vectors.

## Key Examples

- Two vectors  $u$  and  $v$  are linearly dependent if and only if one is a multiple of the other. If they are linearly independent, then they span a plane through the origin. Further, inserting  $w$  into the collection produces a linearly independent set if and only if  $w$  is not in  $\text{Span}\{u, v\}$ .
- A set containing the zero vector is automatically linearly dependent.

## *When Homogenous System has Unique Solution?*

**Fact.** *The columns of matrix  $A$  are linearly independent*

$\iff Ax = 0$  *has only the trivial solution*

$\iff$  *there is no free variable.*

## *Matrix Entries*

**Defn.** For matrix  $A$ , notation  $a_{ij}$  means the entry in row  $i$  and column  $j$  of  $A$ .

## *Matrix Addition and Scalar Multiplication*

**Defn.** Matrix **addition** requires the two matrices have the same dimensions. The sum is defined by adding corresponding entries. Similarly, **scalar multiplication** is defined entry-wise.

For example,

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{bmatrix} + \begin{bmatrix} b_{11} & b_{12} & b_{13} \\ b_{21} & b_{22} & b_{23} \end{bmatrix} = \begin{bmatrix} a_{11} + b_{11} & a_{12} + a_{12} & a_{13} + b_{13} \\ a_{21} + b_{21} & a_{22} + a_{22} & a_{23} + b_{23} \end{bmatrix}$$

# Matrix Transpose

**Defn.** The **transpose** of a matrix  $A$ , denoted  $A^T$ , exchanges rows and columns. That is,  $(A^T)_{ij} = A_{ji}$ .

For example: here is a matrix and its transpose

$$\begin{bmatrix} 3 & 4 & 7 \\ -2 & 5 & -3 \end{bmatrix} \quad \begin{bmatrix} 3 & -2 \\ 4 & 5 \\ 7 & -3 \end{bmatrix}$$

# *Square Matrices*

**Defn.** A **square** matrix has equal number of rows and columns.

# Diagonal Matrices

**Defn.** The **diagonal** of a square matrix runs from top-left to bottom-right. A **diagonal matrix** is a square matrix that has zeros off the diagonal (and might or might not have zeroes on the diagonal).

For example

$$\begin{bmatrix} 3 & 0 & 0 \\ 0 & -1 & 0 \\ 0 & 0 & -\pi \end{bmatrix}$$

## *Matrix Multiplication via Matrix-Vector Mult*

**Defn.** *If matrix  $A$  is  $m \times n$  and matrix  $B$  is  $r \times s$ , then for the **product**  $AB$  to be valid it must be that  $n = r$ . If valid, the product  $AB$  has size  $m \times s$ . The columns of the product are the results of multiplying the first matrix by the columns of the second.*

That is,

$$AB = [Ab_1 \ Ab_2 \ \cdots \ Ab_s]$$

where  $b_j$  is the  $j^{\text{th}}$  column of  $B$ .

## *Example of Matrix Multiplication*

The product of a  $2 \times 3$  and  $3 \times 4$  matrix is a  $2 \times 4$  matrix:

$$\begin{bmatrix} 1 & 2 & -1 \\ 0 & 3 & -2 \end{bmatrix} \begin{bmatrix} 3 & 1 & -1 & 5 \\ -2 & 0 & 3 & -4 \\ 1 & -2 & 2 & -1 \end{bmatrix} = \begin{bmatrix} -2 & 3 & 3 & -2 \\ -8 & 4 & 5 & -10 \end{bmatrix}$$

An example detail: the 3rd column of the result is given by  $-\begin{bmatrix} 1 \\ 0 \end{bmatrix} + 3\begin{bmatrix} 2 \\ 3 \end{bmatrix} + 2\begin{bmatrix} -1 \\ -2 \end{bmatrix} = \begin{bmatrix} 3 \\ 5 \end{bmatrix}$ .

## Formula for Entry in Product

Note

$$(AB)_{ij} = \sum_k a_{ik} b_{kj}$$

That is, to calculate entry in row  $i$  and column  $j$  of the product, look at row  $i$  of the first matrix and column  $j$  of the second matrix; then multiply corresponding entries and add.

$$\begin{bmatrix} \cdot & \cdot & \cdot \\ \boxed{0} & \boxed{3} & \boxed{-2} \end{bmatrix} \begin{bmatrix} \cdot & \cdot & \boxed{-1} & \cdot \\ \cdot & \cdot & \boxed{3} & \cdot \\ \cdot & \cdot & \boxed{2} & \cdot \end{bmatrix} = \begin{bmatrix} \cdot & \cdot & \cdot & \cdot \\ \cdot & \cdot & \boxed{5} & \cdot \end{bmatrix}$$

$(0 \times -1) + (3 \times 3) + (-2 \times 2)$

## *Matrix Multiplication is Associative*

**Fact.** *Brackets don't matter.*

For example  $(AB)C = A(BC)$  (and the one product is valid whenever the other one is).

## *Matrix multiplication is not Commutative*

**Fact.** *Order matters.*

There is no guarantee that (and it is unlikely that)  $AB = BA$ .

Indeed, the one product might be valid when the other one is not.

## *The Identity Matrix*

**Defn.** The **identity matrix**  $I_n$  is the  $n \times n$  diagonal matrix with 1's on the diagonal. (We sometimes write just  $I$ .) Its columns are the vectors  $e_i$ : these have 0's in every position except for a 1 in the  $i^{\text{th}}$  position.

If  $A$  is a square matrix, then  $IA = AI = A$ , where  $I$  is the identity matrix of the same size.

## *Matrix Powers*

We use  $A^p$  to mean the product of  $p$  copies of  $A$ .  
(This needs  $A$  to be square.)

## *Transpose and Products*

**Fact.**  $(AB)^T = B^T A^T$

Note that the order is swapped!

## *Elementary Row Operations Revisited*

**Fact.** *Each elementary row operation is equivalent to multiplying on the left by a matrix called an **elementary matrix**.*

## Matrix Transform

**Defn.** If the matrix  $A$  is  $m \times n$ , then the **matrix transform**  $x \mapsto Ax$  has domain  $\mathbb{R}^n$ , codomain  $\mathbb{R}^m$ , and range some subset of  $\mathbb{R}^m$ .

For example, if  $A$  is  $\begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$ , then the transform maps  $(5, 3)$  to  $(3, 5)$ .

## *The Inverse of a Matrix*

**Defn.** The **inverse** of a square matrix  $A$ , denoted  $A^{-1}$ , is the matrix such that  $AA^{-1} = A^{-1}A = I$ .

**Defn.** The inverse is not guaranteed to exist. If it exists, then  $A$  is **invertible**; otherwise  $A$  is **not invertible** or **singular**.

## *Matrix Equation with Invertible Matrix*

**Fact.** *If matrix  $A$  is invertible, then  $Ax = b$  has unique solution  $x = A^{-1}b$ .*

## *Inverse of a $2 \times 2$ Matrix*

The inverse of a  $2 \times 2$  matrix has formula:

$$\begin{bmatrix} a & b \\ c & d \end{bmatrix}^{-1} = \frac{1}{ad - bc} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$$

The formula also captures when the inverse exists: the matrix is invertible if and only if  $ad - bc \neq 0$ .

## *Obtaining Matrix Inverses by Reduction*

One way to find the inverse is to solve the collection of  $n$  vector equations  $A\mathbf{x} = \mathbf{e}_1, \dots, A\mathbf{x} = \mathbf{e}_n$  (where the  $\mathbf{e}_j$  are the columns of  $I_n$  as before). Equivalently:

**ALGOR** *To find inverse of matrix  $A$ , augment with the identity matrix  $I_n$ , then bring to reduced row echelon form.*

## Formulas

**Fact.** If  $A$  and  $B$  are square matrices of the same size:

(a)  $(A^{-1})^{-1} = A$

(b)  $(AB)^{-1} = B^{-1}A^{-1}$  (Note the reversal!)

(c)  $(A^T)^{-1} = (A^{-1})^T.$

## Characterization of Invertible Matrices

The big theorem.

**Fact.** For an  $n \times n$  matrix  $A$ , the following are **equivalent**:

- ⌞  $A$  is invertible
- ⌞  $A$  has  $n$  pivots
- ⌞  $A$  is row equivalent to  $I_n$
- ⌞  $Ax = 0$  has a unique solution
- ⌞ the columns of  $A$  are linearly independent
- ⌞ the columns of  $A$  span  $\mathbb{R}^n$
- ⌞ the range of transform  $x \mapsto Ax$  is all of  $\mathbb{R}^n$

## *Block-Diagonal Matrices*

**Defn.** A **block-diagonal** matrix is one where all blocks off the diagonal are zero.

**Fact.** A block-diagonal matrix is invertible if and only if all the diagonal blocks are invertible. Moreover, its inverse is the block-diagonal matrix with the inverses of the diagonal blocks.

## *Example Inverse of Block-Diagonal Matrix*

$$\begin{bmatrix} 6 & 0 & 0 \\ 0 & 3 & -5 \\ 0 & -5 & 9 \end{bmatrix}^{-1} = \begin{bmatrix} 1/6 & 0 & 0 \\ 0 & 9/2 & 5/2 \\ 0 & 5/2 & 3/2 \end{bmatrix}$$

## Triangular Matrices

**Defn.** A **lower triangular** matrix is one whose entries above the main diagonal are zero. An **upper triangular** matrix is defined similarly.

For example, a diagonal matrix is both lower and upper triangular.

**Fact.** A square triangular matrix is invertible if and only if every entry on the diagonal is nonzero.