3.2, 3.3

Properties of Transpose

Transpose has higher precedence than multiplication and addition, so

$$AB^T = A(B^T)$$
 and $A + B^T = A + (B^T)$

As opposed to the bracketed expressions

$$(AB)^T$$
 and $(A+B)^T$

Example 1

Let
$$A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 5 & 2 \end{pmatrix}$$
 and $B = \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}$.

Find AB^T , and $(AB)^T$.

$$AB^{T} = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 5 & 2 \end{pmatrix} \begin{pmatrix} 1 & 0 & 1 \\ 1 & 1 & 0 \end{pmatrix}^{T} = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 5 & 2 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 0 & 1 \\ 1 & 0 \end{pmatrix}$$
$$= \begin{pmatrix} 2 & 3 \\ 4 & 7 \end{pmatrix}$$

Whereas $(AB)^T$ is undefined.

Theorem 2 (Properties of Transpose) Given matrices A and B so that the operations can be preformed

- 1. $(A^T)^T = A$
- 2. $(A+B)^T = A^T + B^T$ and $(A-B)^T = A^T B^T$

3.
$$(kA)^T = kA^T$$

4. $(AB)^T = B^T A^T$

Matrix Algebra

Theorem 3 (Algebraic Properties of Matrix Multip

- 1. $(k + \ell)A = kA + \ell A$ (Distributivity of scalar multiplication I)
- 2. k(A + B) = kA + kB (Distributivity of scalar multiplication II)
- 3. A(B+C) = AB + AC (Distributivity of matrix multiplication)
- 4. A(BC) = (AB)C (Associativity of matrix multiplication)
- 5. A + B = B + A (Commutativity of matrix addition)
- 6. (A + B) + C = A + (B + C) (Associativity of matrix addition)
- 7. k(AB) = A(kB) (Commutativity of Scalar Multiplication)

The matrix 0 is the identity of matrix addition. That is, given a matrix A,

A + 0 = 0 + A = A.

Further 0A = A0 = 0, where 0 is the appropriately sized 0 matrix.

Note that it is possible to have two non-zero matrices which multiply to 0.

Example 4

$$\begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} = \begin{pmatrix} 1-1 & 1-1 \\ -1+1 & -1+1 \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

The matrix I is the identity of matrix multiplication. That is, given an $m \times n$ matrix A,

$$AI_n = I_m A = A$$

Theorem 5 If R is in reduced row echelon form then either R = I, or R has a row of zeros.

Theorem 6 (Power Laws) For any square matrix A,

$$A^r A^s = A^{r+s}$$
 and $(A^r)^s = A^{rs}$

1.
$$\begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 1 \\ 2 & 2 & 0 \end{pmatrix}^{4} = \left(\begin{pmatrix} 0 & 0 & 1 \\ 1 & 0 & 1 \\ 2 & 2 & 0 \end{pmatrix}^{2} \right)^{2}$$

2. Find
$$A^6$$
, where

$$A = \left(\begin{array}{cc} \mathbf{1} & \mathbf{0} \\ \mathbf{1} & \mathbf{1} \end{array}\right)$$

$$A^{6} = A^{2}A^{4} = A^{2} \left(A^{2}\right)^{2}.$$

Now $A^{2} = \begin{pmatrix} 1 & 0 \\ 2 & 1 \end{pmatrix}$, so

$$A^{2} \left(A^{2}\right)^{2} = \begin{pmatrix} 1 & 0 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 2 & 1 \end{pmatrix}^{2} = \begin{pmatrix} 1 & 0 \\ 2 & 1 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ 3 & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 1 & 0 \\ 5 & 1 \end{pmatrix}$$

Inverse of a matrix

Given a square matrix A, the *inverse of* A, denoted A^{-1} , is defined to be the matrix such that

$$AA^{-1} = A^{-1}A = I$$

Note that inverses are only defined for square matrices

Note Not all matrices have inverses.

If A has an inverse, it is called *invertible*.

If A is not invertible it is called *singular*.

1.
$$A = \begin{pmatrix} 1 & 2 \\ 2 & 5 \end{pmatrix} \qquad A^{-1} = \begin{pmatrix} 5 & -2 \\ -2 & 1 \end{pmatrix}$$

Check:
$$\begin{pmatrix} 1 & 2 \\ 2 & 5 \end{pmatrix} \begin{pmatrix} 5 & -2 \\ -2 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

2.
$$A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}$$
 Has no inverse

3.
$$A = \begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix}$$
 $A^{-1} = \begin{pmatrix} 3 & -1 & -1 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix}$
Check: $\begin{pmatrix} 1 & 1 & 1 \\ 1 & 2 & 1 \\ 1 & 1 & 2 \end{pmatrix} \begin{pmatrix} 3 & -1 & -1 \\ -1 & 1 & 0 \\ -1 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
4. $A = \begin{pmatrix} 1 & 2 & 1 \\ 2 & 1 & 3 \\ 3 & 3 & 4 \end{pmatrix}$ Has no inverse

Inverses of 2×2 Matrices

Given a 2×2 matrix

$$A = \left(\begin{array}{cc} a & b \\ c & d \end{array}\right)$$

A is invertible if and only if $ad - bc \neq 0$ and

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

The quantity ad - bc is called the *determinant* of the matrix and is written det(A), or |A|.

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 3 \end{pmatrix} \qquad A^{-1} = \frac{1}{-3} \begin{pmatrix} 3 & -2 \\ -3 & 1 \end{pmatrix} = \begin{pmatrix} -1 & \frac{2}{3} \\ 1 & -\frac{1}{3} \end{pmatrix}$$

Check: $\frac{1}{-3} \begin{pmatrix} 3 & -2 \\ -3 & 1 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ 3 & 3 \end{pmatrix} = -\frac{1}{3} \begin{pmatrix} -3 & 0 \\ 0 & -3 \end{pmatrix} = I$

Algebra of Invertibility

Theorem 10 Given an invertible matrix A:

1.
$$(A^{-1})^{-1} = A$$
,

2.
$$(A^n)^{-1} = (A^{-1})^n \quad (=A^{-n}),$$

3.
$$(kA)^{-1} = \frac{1}{k}A^{-1}$$
,

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

Theorem 11 Given two invertible matrices A and B

$$(AB)^{-1} = B^{-1}A^{-1}.$$

Proof: Let A and B be invertible matricies and let C = AB, so $C^{-1} = (AB)^{-1}$.

Consider C = AB. Multiply both sides on the left by A^{-1} :

$$A^{-1}C = A^{-1}AB = B.$$

Multiply both sides on the left by B^{-1} .

$$B^{-1}A^{-1}C = B^{-1}B = I.$$

So, $B^{-1}A^{-1}$ is the matrix you need to multiply Cby to get the identity.

Thus, by the definition of inverse

$$B^{-1}A^{-1} = C^{-1} = (AB)^{-1}.$$

A Method for Inverses

Given a square matrix A and a vector $\mathbf{b} \in \mathbb{R}^n$, consider the equation

$$A\mathbf{x} = \mathbf{b}$$

This represents a system of equations with coefficient matrix A.

Multiply both sides by A^{-1} on the left, to get

$$A^{-1}A\mathbf{x} = A^{-1}\mathbf{b}.$$

But $A^{-1}A = I_n$ and $I\mathbf{x} = \mathbf{x}$, so we have

 $\mathbf{x} = A^{-1}\mathbf{b}$.

Note that we have a unique solution. The assumption that A is invertible is equalvalent to the assumption that $A\mathbf{x} = \mathbf{b}$ has unique solution.

During the course of Gauss-Jordan elimination on the augmented matrix $(A|\mathbf{b})$ we reduce $A \rightarrow I$ and $\mathbf{b} \to A^{-1}\mathbf{b}$, so $(A|\mathbf{b}) \to (I|A^{-1}\mathbf{b})$.

If we instead augment A with I, row reducing will produce (hopefully) I on the left and A^{-1} on the right, so $(A|I) \rightarrow (I|A^{-1})$.

The Method:

- 1. Augment A with I
- 2. Use Gauss-Jordan to obtain $(I|A^{-1})$.
- 3. If I does not appear on the left, A is not invertable.

Otherwise, A^{-1} is given on the right.

1. Find
$$A^{-1}$$
, where

$$A = \left(\begin{array}{rrrr} 1 & 2 & 3 \\ 2 & 5 & 5 \\ 3 & 5 & 8 \end{array}\right)$$

Augment with *I* and row reduce:

$$\begin{pmatrix} 1 & 2 & 3 & | & 1 & 0 & 0 \\ 2 & 5 & 5 & | & 0 & 1 & 0 \\ 3 & 5 & 8 & | & 0 & 0 & 1 \end{pmatrix} \quad \begin{array}{c} R_2 \to R_2 - 2R_1 \\ R_3 \to R_3 - 3R_1 \\ \\ \begin{pmatrix} 1 & 2 & 3 & | & 1 & 0 & 0 \\ 0 & 1 & -1 & | & -2 & 1 & 0 \\ 0 & -1 & -1 & | & -3 & 0 & 1 \end{pmatrix} \quad \begin{array}{c} R_3 \to R_3 + R_2 \\ \\ \begin{pmatrix} 1 & 2 & 3 & | & 1 & 0 & 0 \\ 0 & 1 & -1 & | & -2 & 1 & 0 \\ 0 & 0 & -2 & | & -5 & 1 & 1 \end{pmatrix} \quad \begin{array}{c} R_3 \to -\frac{1}{2}R_3 \\ \\ R_2 \to R_2 + R_3 \\ \end{array}$$

$$\begin{pmatrix} 1 & 2 & 0 & | & -13/2 & 3/2 & 3/2 \\ 0 & 1 & 0 & | & 5/2 & -1/2 & -1/2 \\ 0 & 0 & 1 & | & 5/2 & -1/2 & -1/2 \\ \end{array}) \quad \begin{array}{c} R_1 \to R_1 - 3R_3 \\ R_2 \to R_2 + R_3 \\ \end{array}$$

$$\begin{pmatrix} 1 & 2 & 0 & | & -13/2 & 3/2 & 3/2 \\ 0 & 1 & 0 & | & 1/2 & 1/2 & -1/2 \\ 0 & 0 & 1 & | & 5/2 & -1/2 & -1/2 \\ \end{array}) \quad \begin{array}{c} R_1 \to R_1 - 2R_2 \\ \end{array}$$

3.2, 3.3

So

$$A^{-1} = \frac{1}{2} \begin{pmatrix} -15 & 1 & 5\\ 1 & 1 & -1\\ 5 & -1 & -1 \end{pmatrix}$$

To check inverse multiply together:

$$AA^{-1} = \begin{pmatrix} 1 & 2 & 3 \\ 2 & 5 & 5 \\ 3 & 5 & 8 \end{pmatrix} \frac{1}{2} \begin{pmatrix} -15 & 1 & 5 \\ 1 & 1 & -1 \\ 5 & -1 & -1 \end{pmatrix}$$
$$= \frac{1}{2} \begin{pmatrix} 2 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 2 \end{pmatrix} = I$$

2. Solve $A\mathbf{x} = \mathbf{b}$ in the case where $\mathbf{b} = (2, 2, 4)^T$.

$$\mathbf{x} = A^{-1}\mathbf{b} = \frac{1}{2} \begin{pmatrix} -15 & 1 & 5 \\ 1 & 1 & -1 \\ 5 & -1 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ 2 \\ 4 \end{pmatrix}$$
$$= \frac{1}{2} \begin{pmatrix} -18 \\ 0 \\ 4 \end{pmatrix} = \begin{pmatrix} -9 \\ 0 \\ 2 \end{pmatrix}$$

3. Solve $A\mathbf{x} = \mathbf{b}$ in the case where $\mathbf{b} = (2, 0, 2)^T$.

$$\mathbf{x} = A^{-1}\mathbf{b} = \frac{1}{2} \begin{pmatrix} -15 & 1 & 5 \\ 1 & 1 & -1 \\ 5 & -1 & -1 \end{pmatrix} \begin{pmatrix} 2 \\ 0 \\ 2 \end{pmatrix}$$
$$= \frac{1}{2} \begin{pmatrix} -20 \\ 0 \\ 8 \end{pmatrix} = \begin{pmatrix} -9 \\ 0 \\ 4 \end{pmatrix}$$

4. Give a solution to $A\mathbf{x} = \mathbf{b}$ in the general case where $b = (b_1, b_2, b_3)$

$$\mathbf{x} = \frac{1}{2} \begin{pmatrix} -15 & 1 & 5 \\ 1 & 1 & -1 \\ 5 & -1 & -1 \end{pmatrix} \begin{pmatrix} b_1 \\ b_2 \\ b_3 \end{pmatrix}$$
$$= \frac{1}{2} \begin{pmatrix} -15b_1 + b_2 + 5b_3 \\ b_1 + b_2 - b_3 \\ 5b_1 - b_2 - b_3 \end{pmatrix}$$

Elementary Matrices

Definition 13 An Elementary matrix is a matrix obtained by preforming a single row operation on the identity matrix.

1.
$$\begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
 $(R_1 \to 2R_1)$
2. $\begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$ $(R_2 \to R_2 + 3R_1)$
3. $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$ $(R_1 \leftrightarrow R_2)$

Theorem 15 If E is an elementary matrix obtained from I_m by preforming the row operation R and A is any $m \times n$ matrix, then EA is the matrix obtained by preforming the same row operation Ron A.

Example 16

$$A = \left(\begin{array}{rrrr} 1 & 1 & 1 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{array}\right)$$

1.
$$\begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix} = \begin{pmatrix} 2 & 2 & 2 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix} \sim 2R_2 \text{ on } A$$

2.

$$\begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 \\ 5 & 4 & 3 \\ 3 & 2 & 1 \end{pmatrix} \sim \begin{array}{c} R_2 \to R_2 + 3R_1 \\ \text{on } A \end{array}$$

3.

3.2, 3.3 Inverting Matrices P. Danziger $\begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix} \begin{pmatrix} 1 & 1 & 1 \\ 2 & 1 & 0 \\ 3 & 2 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 1 & 1 \\ 3 & 2 & 1 \\ 2 & 1 & 0 \end{pmatrix} \sim \begin{array}{c} R_2 \leftrightarrow R_3 \\ \text{on } A \end{array}$

Inverses of Elementary Matrices

If E is an elementary matrix then E is invertible and E^{-1} is an elementary matrix corresponding to the row operation that undoes the one that generated E. Specifically:

- If E was generated by an operation of the form $R_i \to cR_i$ then E^{-1} is generated by $R_i \to \frac{1}{c}R_i$.
- If E was generated by an operation of the form $R_i \rightarrow R_i + cR_j$ then E^{-1} is generated by $R_i \rightarrow R_i - cR_j$.
- If E was generated by an operation of the form $R_i \leftrightarrow R_j$ then E^{-1} is generated by $R_i \leftrightarrow R_j$.

1.
$$E = \begin{pmatrix} 2 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

2. $E = \begin{pmatrix} 1 & 0 & 0 \\ 3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$
3. $E = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & 1 & 0 \end{pmatrix}$

$$E^{-1} = \begin{pmatrix} \frac{1}{2} & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
$$E^{-1} = \begin{pmatrix} 1 & 0 & 0 \\ -3 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$
$$E^{-1} = E$$

Elementary Matricies and Solving Equations

Consider the steps of Gauss Jordan elimination to find the solution to a system of equations $A\mathbf{x} = \mathbf{b}$. This consists of a series of row operations, each of which is equivalent to multiplying on the left by an elementary matrix E_i .

$$A \xrightarrow{\text{Ele. row ops.}} B,$$

Where B is the RREF of A.

So $E_k E_{k-1} \dots E_2 E_1 A = B$ for some appopriately defined elementary matrices $E_1 \dots E_k$.

Thus $A = E_1^{-1} E_2^{-1} \dots E_{k-1}^{-1} E_k^{-1} B$

Now if
$$B = I$$
 (so the RREF of *A* is *I*), then
 $A = E_1^{-1}E_2^{-1} \dots E_{k-1}^{-1}E_k^{-1}$

and $A^{-1} = E_k E_{k-1} \dots E_2 E_1$

Theorem 18 *A* is invertable if and only if it is the product of elementary matrices.

3.2, 3.3

Summing Up Theorem

Theorem 19 (Summing up Theorem Version 1) For any square $n \times n$ matrix A, the following are equivalent statements:

- 1. A is invertible.
- 2. The RREF of A is the identity, I_n .
- 3. The equation $A\mathbf{x} = \mathbf{b}$ has unique solution (namely $\mathbf{x} = A^{-1}\mathbf{b}$).
- 4. The homogeneous system Ax = 0 has only the trivial solution (x = 0)
- 5. The REF of A has exactly n pivots.
- 6. A is the product of elementary matrices.