

## Computational Foundations of Cognitive Science

## Lecture 12: Inverses; Matrices with Special Forms

Frank Keller

School of Informatics  
University of Edinburgh  
keller@inf.ed.ac.uk

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Reading: Anton and Busby, Chs. 3.2, 3.6



## Definition of the Inverse

Every real number  $a$  has a reciprocal  $a^{-1} = 1/a$  such that  $a \cdot a^{-1} = 1$ . The analogue for a matrix  $A$  is its *inverse*  $A^{-1}$ .

## Definition: Matrix Inverse

If  $A$  is a square matrix, and if there is a matrix  $B$  with the same size such that  $AB = I$ , then  $A$  is invertible (non-singular), and  $B$  is the inverse of  $A$ . If there is no such matrix  $B$ , then  $A$  is singular.

Note that if  $AB = I$ , then also  $BA = I$ , i.e.,  $B$  is also invertible and an inverse of  $A$ .

## Example

$$A = \begin{bmatrix} 2 & -5 \\ -1 & 3 \end{bmatrix} \quad B = \begin{bmatrix} 3 & 5 \\ 1 & 2 \end{bmatrix} \quad AB = \begin{bmatrix} 2 & -5 \\ -1 & 3 \end{bmatrix} \begin{bmatrix} 3 & 5 \\ 1 & 2 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} = I$$



## Definition of the Inverse

In general, a square matrix with a row or column of zeros is singular, i.e., it has no inverse.

The row or column of zeros causes the result of matrix multiplication to also have a row or column of zeros, i.e., it can't be  $I$ .

## Example

$$\text{Assume } A = \begin{bmatrix} 1 & 4 & 0 \\ 2 & 5 & 0 \\ 3 & 6 & 0 \end{bmatrix} = [c_1 \quad c_2 \quad \mathbf{0}].$$

Then apply the definition of matrix multiplication to compute  $BA = B [c_1 \quad c_2 \quad \mathbf{0}] = [Bc_1 \quad Bc_2 \quad \mathbf{0}]$ . This shows that  $BA \neq I$ .



## Definition of the Inverse

## Theorem: Uniqueness of the Inverse

If  $A$  is an invertible matrix, and if  $B$  and  $C$  are both inverses of  $A$ , then  $B = C$ , i.e., an invertible matrix has a unique inverse.

The uniqueness of the inverse implies  $AA^{-1} = I$  and  $A^{-1}A = I$ .

Finding the inverse of a matrix is generally non-trivial. However, for  $2 \times 2$  matrices, there is a simple formula. For this we need the notion of a **determinant** of a matrix.

## Definition: Determinant of a Matrix

The determinant of a  $2 \times 2$  matrix  $A$  is given by:

$$\det(A) = \begin{vmatrix} a & b \\ c & d \end{vmatrix} = ad - bc$$

## Computation of the Inverse

## Theorem: Computation of the Inverse

The matrix  $A = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$  is invertible iff  $\det(A) \neq 0$ , in which case:

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

## Examples

$$A = \begin{bmatrix} 6 & 1 \\ 5 & 2 \end{bmatrix} \quad B = \begin{bmatrix} -1 & 2 \\ 3 & -6 \end{bmatrix}$$

$$\det(A) = 6 \cdot 2 - 1 \cdot 5 = 7 \quad \det(B) = (-1)(-6) - 2 \cdot 3 = 0$$

$$A^{-1} = \frac{1}{7} \begin{bmatrix} 2 & -1 \\ -5 & 6 \end{bmatrix} = \begin{bmatrix} \frac{2}{7} & -\frac{1}{7} \\ -\frac{5}{7} & \frac{6}{7} \end{bmatrix}$$

## Properties of the Inverse

## Theorem: Product of Inverses; Transpose

If  $A$  and  $B$  are invertible matrices of the same size, then  $AB$  is invertible and  $(AB)^{-1} = B^{-1}A^{-1}$ .

If  $A$  is invertible, then  $(A^T)^{-1} = (A^{-1})^T$ .

## Example

$$A = \begin{bmatrix} 1 & 2 \\ 1 & 3 \end{bmatrix} \quad B = \begin{bmatrix} 3 & 2 \\ 2 & 2 \end{bmatrix} \quad AB = \begin{bmatrix} 7 & 6 \\ 9 & 8 \end{bmatrix} \quad (AB)^{-1} = \begin{bmatrix} 4 & -3 \\ -\frac{9}{2} & \frac{7}{2} \end{bmatrix}$$

$$A^{-1} = \begin{bmatrix} 3 & -2 \\ -1 & 1 \end{bmatrix} \quad B^{-1} = \begin{bmatrix} 1 & -1 \\ -1 & \frac{3}{2} \end{bmatrix} \quad B^{-1}A^{-1} = \begin{bmatrix} 4 & -3 \\ -\frac{9}{2} & \frac{7}{2} \end{bmatrix}$$

In general, a product of any number of invertible matrices is invertible, and the inverse of the product is the product of the inverses in the reverse order.

## Diagonal Matrices

A square matrix in which all the entries off the main diagonal are zero is a **diagonal matrix**:

$$D = \begin{bmatrix} d_1 & 0 & \cdots & 0 \\ 0 & d_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & d_n \end{bmatrix}$$

A diagonal matrix is invertible iff all the  $d_i$  are non-zero in which case the inverse is

$$D^{-1} = \begin{bmatrix} 1/d_1 & 0 & \cdots & 0 \\ 0 & 1/d_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1/d_n \end{bmatrix}$$

## Diagonal Matrices

Matrix products involving diagonal matrices are easy to compute:

### Examples

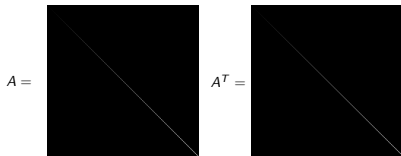
$$\begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ a_{21} & a_{22} & a_{23} & a_{24} \\ a_{31} & a_{32} & a_{33} & a_{34} \end{bmatrix} = \begin{bmatrix} d_1 a_{11} & d_1 a_{12} & d_1 a_{13} & d_1 a_{14} \\ d_2 a_{21} & d_2 a_{22} & d_2 a_{23} & d_2 a_{24} \\ d_3 a_{31} & d_3 a_{32} & d_3 a_{33} & d_3 a_{34} \end{bmatrix}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \end{bmatrix} \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} = \begin{bmatrix} d_1 a_{11} & d_2 a_{12} & d_3 a_{13} \\ d_1 a_{21} & d_2 a_{22} & d_3 a_{23} \\ d_1 a_{31} & d_2 a_{32} & d_3 a_{33} \\ d_1 a_{41} & d_2 a_{42} & d_3 a_{43} \end{bmatrix}$$



## Example: Representing Images

Let us again represent matrices as greyscale images. Let  $A$  be a diagonal matrix with the values  $1 \dots 200$  on the diagonal:



Transposing a diagonal matrix doesn't change it.



## Mid-lecture Problem

Indicate whether the following statements are true or false. Justify your answers.

- 1 If  $A$  is invertible, then so is  $I - A$ .
- 2 If  $A$  is a diagonal matrix, then so is  $AA^T$ .
- 3 If  $A$  and  $B$  are diagonal matrices, then  $AB = BA$ .



## Triangular Matrices

A square matrix is *triangular* if all entries to the left of the main diagonal are zero (upper triangular), or if all entries to the right of main diagonal are zero (lower triangular). More formally:

- $A = [a_{ij}]$  is upper triangular iff  $a_{ij} = 0$  for all  $i > j$ ;
- $A = [a_{ij}]$  is lower triangular iff  $a_{ij} = 0$  for all  $i < j$ .

Matrix can be both upper and lower triangular (e.g., diagonal matrix).

### Example

A  $4 \times 4$  upper and lower triangular matrix has the form:

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} \\ 0 & a_{22} & a_{23} & a_{24} \\ 0 & 0 & a_{33} & a_{34} \\ 0 & 0 & 0 & a_{44} \end{bmatrix} \begin{bmatrix} a_{11} & 0 & 0 & 0 \\ a_{21} & a_{22} & 0 & 0 \\ a_{31} & a_{23} & a_{33} & 0 \\ a_{41} & a_{24} & a_{43} & a_{44} \end{bmatrix}$$



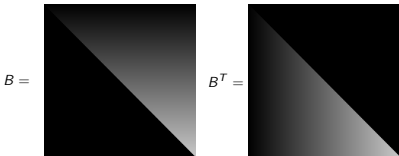
## Triangular Matrices

### Theorem: Properties of Triangular Matrices

- The transpose of a lower (upper) triangular matrix is upper (lower) triangular.
- A product of lower (upper) triangular matrices is lower (upper) triangular.
- A triangular matrix is invertible iff its diagonal entries are all non-zero.
- The inverse of a lower (upper) triangular matrix is lower (upper) triangular.

## Example: Representing Images

Now let  $B$  be an upper triangular matrix with the values  $1 \dots 200$  on the rows  $1 \dots 200$ :



Transposing an upper triangular matrix yields a lower triangular matrix.

## Triangular Matrices

### Example

$$A = \begin{bmatrix} 1 & 3 & -1 \\ 0 & 2 & 4 \\ 0 & 0 & 5 \end{bmatrix} \quad B = \begin{bmatrix} 3 & -1 & 2 \\ 0 & 0 & -1 \\ 0 & 0 & 1 \end{bmatrix}$$

$A$  is invertible but  $B$  isn't. Also,  $A^{-1}$ ,  $AB$  and  $BA$  must be upper triangular:

$$A^{-1} = \begin{bmatrix} 1 & -\frac{3}{2} & \frac{7}{5} \\ 0 & \frac{1}{2} & -\frac{2}{5} \\ 0 & 0 & \frac{1}{5} \end{bmatrix} \quad AB = \begin{bmatrix} 3 & -2 & -2 \\ 0 & 0 & 2 \\ 0 & 0 & 5 \end{bmatrix} \quad BA = \begin{bmatrix} 3 & 5 & -1 \\ 0 & 0 & -5 \\ 0 & 0 & 5 \end{bmatrix}$$

## Symmetric Matrices

A square matrix  $A$  is *symmetric* if  $A^T = A$  and *skew-symmetric* if  $A^T = -A$ . More formally:

- $A = [a_{ij}]$  symmetric iff  $(A)_{ij} = (A)_{ji}$ , or equivalently,  $a_{ij} = a_{ji}$ ;
- $A = [a_{ij}]$  skew-symmetric iff  $(A)_{ij} = -(A)_{ji}$ , or equivalently,  $a_{ij} = -a_{ji}$ .

### Examples

$$\begin{bmatrix} 7 & -3 \\ -3 & 5 \end{bmatrix} \quad \begin{bmatrix} 1 & 4 & 5 \\ 4 & -3 & -6 \\ 5 & -6 & 7 \end{bmatrix} \quad \begin{bmatrix} 0 & 3 \\ -3 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & -4 & 5 \\ 4 & 0 & -9 \\ -5 & 9 & 0 \end{bmatrix}$$

## Symmetric Matrices

### Theorem: Properties of Symmetric Matrices

If  $A$  and  $B$  are symmetric matrices with the same size, and  $k$  is a scalar, then:

- $A^T$  is symmetric
- $A + B$  and  $A - B$  are symmetric
- $kA$  is symmetric
- $AB$  is symmetric iff  $A$  and  $B$  commute, i.e., iff  $AB = BA$
- if  $A$  is symmetric and invertible, then  $A^{-1}$  is symmetric

### Examples

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

## Example: Representing Images

We get a symmetric matrix  $C$  by adding  $B$  and  $B^T$ :

$$C = \begin{bmatrix} \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \end{bmatrix} \quad C^T = \begin{bmatrix} \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \\ \text{img} \end{bmatrix}$$

Transposing a symmetric matrix doesn't change it.

## Matrices of the form $AA^T$ and $A^T A$

If  $A$  is an  $m \times n$  matrix, then  $A^T$  is an  $n \times m$  matrix, which means that  $AA^T$  and  $A^T A$  are square matrices of size  $m \times m$  and  $n \times n$ .

$AA^T$  and  $A^T A$  are always symmetric, since:

$$(AA^T)^T = (A^T)^T A^T = AA^T \quad \text{and} \quad (A^T A)^T = A^T (A^T)^T = A^T A.$$

### Example

$$A^T A = \begin{bmatrix} \mathbf{a}_1^T \\ \mathbf{a}_2^T \\ \vdots \\ \mathbf{a}_n^T \end{bmatrix} \begin{bmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \cdots & \mathbf{a}_n \end{bmatrix} = \begin{bmatrix} \mathbf{a}_1^T \mathbf{a}_1 & \mathbf{a}_1^T \mathbf{a}_2 & \cdots & \mathbf{a}_1^T \mathbf{a}_n \\ \mathbf{a}_2^T \mathbf{a}_1 & \mathbf{a}_2^T \mathbf{a}_2 & \cdots & \mathbf{a}_2^T \mathbf{a}_n \\ \vdots & \vdots & \ddots & \vdots \\ \mathbf{a}_n^T \mathbf{a}_1 & \mathbf{a}_n^T \mathbf{a}_2 & \cdots & \mathbf{a}_n^T \mathbf{a}_n \end{bmatrix}$$

By definition of the inner product,

$$\mathbf{a}_1^T \mathbf{a}_2 = \mathbf{a}_1 \cdot \mathbf{a}_2 = \mathbf{a}_2 \cdot \mathbf{a}_1 = \mathbf{a}_2^T \mathbf{a}_1, \text{ hence the matrix is symmetric.}$$

## Summary

- The inverse of a matrix  $A$  is  $A^{-1}$  with  $AA^{-1} = I$ ;
- the determinant of a  $2 \times 2$  matrix is  $\det(A) = ad - bc$ ;
- $A^{-1} = \frac{1}{\det(A)} \begin{bmatrix} d & -b \\ -c & a \end{bmatrix}$ ;
- diagonal matrix: all entries off the main diagonal are zero;
- triangular matrix: all entries to the left or right of the main diagonal are zero;
- symmetric matrix:  $A^T = A$ ;
- $AA^T$  and  $A^T A$  are always symmetric.