

# AI511/MM505 Linear Algebra with Applications

Lecture 12 – Eigenvalues and Eigenvectors (cont.),  
Diagonalisation

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# Outline

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Recap

Eigenvalues and eigenvectors (cont.)

Complex eigenvalues

Eigenspaces and geometric multiplicity

Diagonalisation

## Recap

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## Definition 3.1.3 of Johnston (2021)

We say that two matrices  $A, D \in \mathbb{R}^{n \times n}$  are similar if there exists an invertible  $P \in \mathbb{R}^{n \times n}$  such that  $A = PDP^{-1}$ .

# Eigenvalues and eigenvectors

## Definition 3.3.1 of Johnston (2021)

Suppose  $A$  is a square matrix. A non-zero vector  $v$  is called an *eigenvector* of  $A$  if there is a scalar  $\lambda$  such that

$$Av = \lambda v.$$

Such a scalar  $\lambda$  is called the *eigenvalue* of  $A$  corresponding to  $v$ .

To find the eigenvalues and eigenvectors of a matrix,

- (i) we first look for the eigenvalues by solving  $\det(A - \lambda I) = 0$ ;
- (ii) once we know the eigenvalues, we find the eigenvectors by solving the homogeneous system of linear equations

$$(A - \lambda I)v = 0$$

for each eigenvalue  $\lambda$ .

## Example

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- The eigenvalues of the matrix

$$A = \begin{bmatrix} 1 & 2 \\ 5 & 4 \end{bmatrix}$$

are  $\lambda_1 = -1$  and  $\lambda_2 = 6$ .

- The eigenvectors corresponding to the eigenvalue  $\lambda_1 = -1$  are the vectors of the form  $v = c(-1, 1)$ , where  $v \in \mathbb{R}^2$ ,  $c \in \mathbb{R}$ , and  $c \neq 0$ .
- The eigenvectors corresponding to the eigenvalue  $\lambda_1 = 6$  are the vectors of the form  $v = c(2/5, 1)$ , where  $v \in \mathbb{R}^2$ ,  $c \in \mathbb{R}$ , and  $c \neq 0$ .

# Characteristic polynomial

## Definition 3.3.2 of Johnston (2021)

Suppose  $A$  is a square matrix. The function  $p_A : \mathbb{R} \rightarrow \mathbb{R}$  defined by

$$p_A(\lambda) = \det(A - \lambda I)$$

is called the *characteristic polynomial* of  $A$ .

## Definition 3.3.3 of Johnston (2021)

Suppose that  $A \in \mathbb{R}^{n \times n}$  with eigenvalue  $\lambda$ . The *algebraic multiplicity* of  $\lambda$  is its multiplicity as a root of  $A$ 's characteristic polynomial.

## Eigenvalues and eigenvectors (cont.)

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## Eigenvalues and eigenvectors (cont.)

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Complex eigenvalues

## No real eigenvalues

- The characteristic polynomial of the matrix

$$A = \begin{bmatrix} -3 & -2 \\ 4 & 1 \end{bmatrix}$$

is given by

$$p_A(\lambda) = \lambda^2 + 2\lambda + 5.$$

- If we restrict our attention to  $\lambda \in \mathbb{R}$ , the quadratic equation

$$\lambda^2 + 2\lambda + 5 = 0$$

has no solutions (we have that  $\Delta = b^2 - 4ac = -16 < 0$ ) which in turn would mean that matrix  $A$  has no eigenvalues.

# Fundamental theorem of algebra

## Theorem A.2.3 of Johnston (2021)

Every non-constant polynomial has at least one complex root.

- Hence, if we allow for complex eigenvalues, every matrix has at least one eigenvalue (which might be complex).
- In fact, every  $n \times n$  matrix has *exactly*  $n$  complex eigenvalues, counting algebraic multiplicity.
- We need to specify that a matrix has exactly  $n$  complex eigenvalues counting algebraic multiplicity even if the entries of a matrix are real as we saw in the example.

# Eigenvalues of symmetric matrices

A special case of Theorem 3.3.2 of Johnston (2021)

If  $A \in \mathbb{R}^{n \times n}$  is symmetric, then all of its eigenvalues are real.

## Eigenvalues and eigenvectors (cont.)

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Eigenspaces and geometric multiplicity

# Eigenspace

The set of all eigenvectors of a matrix  $A \in \mathbb{R}^{n \times n}$  corresponding to a particular eigenvalue  $\lambda$  (together with the zero vector) is  $\text{null}(A - \lambda I)$ .

## Definition 3.3.4 of Johnston (2021)

Suppose  $A$  is a square matrix with eigenvalue  $\lambda$ . The set of all eigenvectors of  $A$  corresponding to  $\lambda$ , together with the zero vector, is called the *eigenspace* of  $A$  corresponding to  $\lambda$ .

Instead of finding all eigenvectors corresponding to a particular eigenvalue, we typically just find a basis of that eigenspace.

## Blackboard example: eigenspaces

- Compute bases of the eigenspaces of the matrix

$$A = \begin{bmatrix} 2 & 0 & -3 \\ 1 & -1 & -1 \\ 0 & 0 & -1 \end{bmatrix}.$$

- Recall that the eigenvalues of  $A$  are  $2$  (with algebraic multiplicity  $1$ ) and  $-1$  (with algebraic multiplicity  $2$ ).
- Geometrically, this means that the eigenspace corresponding to  $\lambda = 2$  is a line, while the eigenspace corresponding to  $\lambda = -1$  is a plane.
- Every vector on that line is stretched by a factor of  $\lambda = 2$ , and every vector in that plane is reflected through the origin.

## Definition 3.3.5 of Johnston (2021)

Suppose  $A$  is a square matrix with eigenvalue  $\lambda$ . The geometric multiplicity of  $\lambda$  is the dimension of its corresponding eigenspace.

In other words, the geometric multiplicity of the eigenvalue  $\lambda$  of  $A$  is  $\text{nullity}(A - \lambda I)$ .

## Geometric multiplicity cannot exceed algebraic multiplicity

### Theorem 3.3.3 of Johnston (2021)

For each eigenvalue of a square matrix, the geometric multiplicity is less than or equal to the algebraic multiplicity.

## Blackboard example: smaller geometric multiplicity

Compute the eigenvalues and their algebraic as well as geometric multiplicities of the matrix

$$A = \begin{bmatrix} 2 & 2 & 3 \\ 0 & -1 & 0 \\ 0 & 1 & -1 \end{bmatrix}.$$

# Diagonalisation

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# Diagonalisable matrices

## Definition 3.4.1 of Johnston (2021)

A matrix  $A \in \mathbb{R}^{n \times n}$  is called diagonalisable if there is a diagonal matrix  $D \in \mathbb{R}^{n \times n}$  and an invertible matrix  $P \in \mathbb{R}^{n \times n}$  such that  $A = PDP^{-1}$ .

## Theorem 3.4.1 of Johnston (2021)

Let  $A \in \mathbb{R}^{n \times n}$  and suppose  $P, D \in \mathbb{R}^{n \times n}$  are such that  $P$  is invertible and  $D$  is diagonal. Then  $A = PDP^{-1}$  if and only if the columns of  $P$  are eigenvectors of  $A$  whose corresponding eigenvalues are the diagonal entries of  $D$  in the same order.

## Blackboard example: diagnosable matrix

- Diagonalise the matrix

$$A = \begin{bmatrix} 1 & 2 \\ 5 & 4 \end{bmatrix}.$$

Recall that the eigenvalues of  $A$  are  $\lambda_1 = -1$  and  $\lambda_2 = 6$  while the corresponding eigenspaces are spanned by  $\{(-1, 1)\}$  and  $\{(2, 5)\}$ .

- How would we find the inverse of  $A$  once we have  $P$ ,  $D$ , and  $P^{-1}$ ?

## Not all matrices are diagonalisable

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- If  $A \in \mathbb{R}^{n \times n}$ , it may not be possible to construct  $D$  and  $P$  so that they have real entries, since  $A$  might have complex (non-real) eigenvalues and eigenvectors.
- Even if we allow for complex eigenvalues and eigenvectors, there may not be a way to choose eigenvectors so that the matrix  $P$  (whose columns are eigenvectors) is invertible.

# Characterisation of diagonalisability

## Theorem 3.4.3 of Johnston (2021)

Suppose  $A \in \mathbb{R}^{n \times n}$ . The following are equivalent:

- (a)  $A$  is diagonalisable over  $\mathbb{R}$ ;
- (b) there exists a basis of  $\mathbb{R}^n$  consisting of eigenvectors of  $A$  (this basis is called an eigenbasis of  $A$ );
- (c) the set of eigenvectors of  $A$  spans all of  $\mathbb{R}^n$ ;
- (d) the sum of the geometric multiplicities of the real eigenvalues of  $A$  is  $n$ .

## Corollary

If  $A \in \mathbb{R}^{n \times n}$  has  $n$  distinct eigenvalues then it is diagonalisable.

Keep in mind that the above corollary only works in one direction.

## Blackboard example: matrix that is not diagonalisable

Show that the matrix

$$A = \begin{bmatrix} 3 & -1 \\ 1 & 1 \end{bmatrix}$$

cannot be diagonalised.

# References

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-  Johnston, Nathaniel (2021). *Introduction to linear and matrix algebra*. Springer Cham.