



NTNU  
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**TMA4115 - Calculus 3**  
**Lecture 21, April 3**

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# Review of last week's lecture

Last week we introduce and studied

- the *dimension* of a vector space,
- the *rank* of a matrix,
- *Markov chains*.



# Today's lecture

Today we shall introduce and study

- *eigenvectors*, *eigenvalues* and *eigenspaces* of square matrices,
- the *characteristic polynomial* of a square matrix.



# Eigenvectors and eigenvalues of square matrices

Let  $A$  be an  $n \times n$  matrix.

- An *eigenvector* of  $A$  is a nonzero vector  $\mathbf{x}$  such that  $A\mathbf{x} = \lambda\mathbf{x}$  for some scalar  $\lambda$ .
- An *eigenvalue* of  $A$  is a scalar  $\lambda$  such that the equation  $A\mathbf{x} = \lambda\mathbf{x}$  has a nontrivial solution.
- If  $\lambda$  is an eigenvalue and  $\mathbf{x}$  is an eigenvector such that  $A\mathbf{x} = \lambda\mathbf{x}$ , then  $\mathbf{x}$  is called an *eigenvector (of  $A$ ) corresponding to  $\lambda$*  and  $\lambda$  is called the *eigenvalue (of  $A$ ) corresponding to  $\mathbf{x}$* .



# Example

Consider the matrix  $A = \begin{bmatrix} 3 & -2 \\ 2 & -2 \end{bmatrix}$ .

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

so 2 is an eigenvalue of  $A$ , and  $\begin{bmatrix} 2 \\ 1 \end{bmatrix}$  is an eigenvector of  $A$  corresponding to the eigenvalue 2.

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

so  $-1$  is an eigenvalue of  $A$ , and  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$  is an eigenvector of  $A$  corresponding to the eigenvalue  $-1$ .



# Eigenspaces

Let  $A$  be an  $n \times n$  matrix and let  $\lambda$  be an eigenvalue of  $A$ .

- A vector  $\mathbf{x} \in \mathbb{R}^n$  is an eigenvector of  $A$  corresponding to  $\lambda$  if and only if  $\mathbf{x}$  is a nontrivial solution of the equation  $(A - \lambda I_n)\mathbf{x} = \mathbf{0}$ .
- The set of all solutions of the equation  $(A - \lambda I_n)\mathbf{x} = \mathbf{0}$  (i.e.,  $\text{Nul}(A - \lambda I_n)$ ) is called the *eigenspace (of  $A$ ) corresponding to  $\lambda$* .
- Notice that an eigenspace of  $A$  is a subspace of  $\mathbb{R}^n$ .
- Notice also that  $\mathbf{0}$  belongs to any eigenspace, even though it is not an eigenvector.



# Example

Consider again the matrix  $A = \begin{bmatrix} 3 & -2 \\ 2 & -2 \end{bmatrix}$ . We have seen that 2 and  $-1$  are eigenvalues of  $A$ . Let us find the corresponding eigenspaces.

The equation  $A\mathbf{x} = 2\mathbf{x}$  is equivalent to the equation  $(A - 2I_2)\mathbf{x} = \mathbf{0}$ . To solve the latter equation, we reduce the matrix  $(A - 2I_2)$  its reduced echelon form.

$$\begin{bmatrix} 1 & -2 \\ 2 & -4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -2 \\ 0 & 0 \end{bmatrix}.$$

We see that the general solution to the equation

$(A - 2I_2)\mathbf{x} = \mathbf{0}$  is  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 2x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 2 \\ 1 \end{bmatrix}$  where  $x_2$  is a free parameter.



## Example (cont.)

It follows that the eigenspace of  $A$  corresponding to the eigenvalue 2 is  $\text{Span} \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right\}$ .

Similar, the equation  $A\mathbf{x} = -1\mathbf{x}$  is equivalent to the equation  $(A + I_2)\mathbf{x} = \mathbf{0}$ . To solve the latter equation, we reduce the matrix  $(A + I_2)$  its reduced echelon form.

$$\begin{bmatrix} 4 & -2 \\ 2 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1/2 \\ 2 & -1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1/2 \\ 0 & 0 \end{bmatrix}.$$

We see that the general solution to the equation  $(A + I_2)\mathbf{x} = \mathbf{0}$  is  $\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}x_2 \\ x_2 \end{bmatrix} = x_2 \begin{bmatrix} 1/2 \\ 1 \end{bmatrix}$  where  $x_2$  is a free parameter.





# Example (cont.)

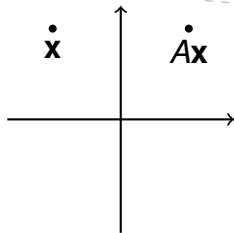
It follows that the eigenspace of  $A$  corresponding to the eigenvalue  $-1$  is  $\text{Span} \left\{ \begin{bmatrix} 1/2 \\ 1 \end{bmatrix} \right\}$ .



# Example

Let  $A = \begin{bmatrix} -1 & 0 \\ 0 & 1 \end{bmatrix}$ . Then the linear transformation  $\mathbf{x} \rightarrow A\mathbf{x}$  sends a vector  $\mathbf{x}$  to the vector which corresponds to the point we get by reflecting the point corresponding to  $\mathbf{x}$  in the  $x_2$ -axis.

It follows that  $-1$  is an eigenvalue of  $A$  and that the corresponding eigenspace is the  $x_1$ -axis, and that  $1$  is an eigenvalue of  $A$  and that the corresponding eigenspace is the  $x_2$ -axis.



# Linearly independent eigenvectors

## Theorem 2

If  $\mathbf{v}_1, \dots, \mathbf{v}_r$  are eigenvectors that correspond to distinct eigenvalues  $\lambda_1, \dots, \lambda_r$  of an  $n \times n$  matrix  $A$ , then the set  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly independent.



# Proof of Theorem 2

Assume, for contradiction, that  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly dependent. Then there is a  $1 \leq p < r$  and such that  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  is linearly independent and

$\mathbf{v}_{p+1} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_p\mathbf{v}_p$  for some scalars  $c_1, c_2, \dots, c_p$ .

Then

$$\lambda_{p+1}\mathbf{v}_{p+1} = A\mathbf{v}_{p+1} = A(c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p) = c_1\lambda_1\mathbf{v}_1 + \dots + c_p\lambda_p\mathbf{v}_p$$

and

$$\lambda_{p+1}\mathbf{v}_{p+1} = \lambda_{p+1}(c_1\mathbf{v}_1 + \dots + c_p\mathbf{v}_p) = c_1\lambda_{p+1}\mathbf{v}_1 + \dots + c_p\lambda_{p+1}\mathbf{v}_p$$

so

$$\mathbf{0} = c_1(\lambda_1 - \lambda_{p+1})\mathbf{v}_1 + \dots + c_p(\lambda_p - \lambda_{p+1})\mathbf{v}_p.$$



# Proof of Theorem 2

Since  $\{\mathbf{v}_1, \mathbf{v}_2, \dots, \mathbf{v}_p\}$  is linearly independent and  $\lambda_k - \lambda_{p+1} \neq 0$  for each  $k = 1, \dots, p$ , it follows that  $c_1 = c_2 = \dots = c_p = 0$ . But then

$\mathbf{v}_{p+1} = c_1\mathbf{v}_1 + c_2\mathbf{v}_2 + \dots + c_p\mathbf{v}_p = \mathbf{0}$  which contradicts the fact that  $\mathbf{v}_{p+1}$  is an eigenvector. So it must be the case that  $\{\mathbf{v}_1, \dots, \mathbf{v}_r\}$  is linearly independent.



# Finding a basis of an eigenspace

Let  $A$  be an  $n \times n$  matrix and let  $\lambda$  be an eigenvalue of  $A$ . The eigenspace of  $A$  corresponding to  $\lambda$  is  $\text{Nul}(A - \lambda I_n)$ . So we can find a basis for the eigenspace of  $A$  corresponding to  $\lambda$  by:

- 1 row reducing  $A - \lambda I_n$  to its reduced echelon form,
- 2 write the solutions to the equation  $(A - \lambda I_n)\mathbf{x} = \mathbf{0}$  (which is equivalent to the equation  $A\mathbf{x} = \mathbf{0}$ ) as a linear combinations of vectors using the free variables as parameters.
- 3 Then these vectors form a basis of the eigenspace of  $A$  corresponding to  $\lambda$ .



# Example

$$\text{Let } A = \begin{bmatrix} 4 & -1 & 6 \\ 2 & 1 & 6 \\ 2 & -1 & 8 \end{bmatrix}.$$

Let us determine if 2 is an eigenvalue of  $A$ , and let us find a basis for the corresponding eigenspace if it is.



# Solution

2 is an eigenvalue of  $A$  if and only if the equation  $(A - 2I_3)\mathbf{x} = \mathbf{0}$  has a nontrivial solution in which case the solution set of the equation is the eigenspace of  $A$  corresponding to 2. We find the solution set of the equation  $(A - 2I_3)\mathbf{x} = \mathbf{0}$  by reducing  $A - 2I_3$  to its reduced echelon form.

$$A - 2I_3 = \begin{bmatrix} 2 & -1 & 6 \\ 2 & -1 & 6 \\ 2 & -1 & 6 \end{bmatrix} \rightarrow \begin{bmatrix} 2 & -1 & 6 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1/2 & 3 \\ 0 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}.$$





# Solution (cont.)

We see that the general solution to the equation

$(A - 2I_3)\mathbf{x} = \mathbf{0}$  is

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} \frac{1}{2}x_2 - 3x_3 \\ x_2 \\ x_3 \end{bmatrix} = x_2 \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix} + x_3 \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \text{ where } x_2$$

and  $x_3$  are free parameters.

It follows that 2 is an eigenvalue of  $A$ , and that

$\left\{ \begin{bmatrix} 1/2 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -3 \\ 0 \\ 1 \end{bmatrix} \right\}$  is a basis for eigenspace of  $A$

corresponding to the eigenvalue 2.



# Finding the eigenvalues of a matrix

Let  $A$  be an  $n \times n$  matrix.

- A scalar  $\lambda$  is an eigenvalue of  $A$  if and only if  $\det(A - \lambda I_n) = 0$ .
- The equation  $\det(A - \lambda I_n) = 0$  is called the *characteristic equation* of  $A$ .
- If we regard  $\lambda$  as an independent variable, then  $\det(A - \lambda I_n)$  is a polynomial of degree  $n$  in  $\lambda$ .
- $\det(A - \lambda I_n)$  is called the *characteristic polynomial* of  $A$ .
- The eigenvalues of  $A$  are the zeros of the characteristic polynomial  $\det(A - \lambda I_n)$  of  $A$ .



# Example

Let us find the eigenvalues of the matrix  $A = \begin{bmatrix} 3 & 0 \\ 4 & -2 \end{bmatrix}$ .

The characteristic polynomial of  $A$  is

$$\det(A - \lambda I_2) = \begin{vmatrix} 3 - \lambda & 0 \\ 4 & -2 - \lambda \end{vmatrix} = (3 - \lambda)(-2 - \lambda) = \lambda^2 - \lambda - 6.$$

The solutions of the characteristic equation  $\lambda^2 - \lambda - 6 = 0$  are 3 and  $-2$ . So the eigenvalues of  $A$  are 3 and  $-2$ .



# Example

In a certain region, 5% of a city's population moves to the surrounding suburbs each year, and 3% of the suburban population moves into the city. In 2000, there were 600,000 residents in the city and 400,000 in the suburbs.

Let us try to find a formula for the number of the people in the city and the number of people in the suburbs for each year.



# Solution (cont.)

Let  $\mathbf{x} = \begin{bmatrix} \text{the population in the city in year } 2000 + i \\ \text{the population in the suburbs in year } 2000 + i \end{bmatrix}$

and  $A = \begin{bmatrix} 0.95 & 0.03 \\ 0.05 & 0.97 \end{bmatrix}$ . Then  $\mathbf{x}_{i+1} = A\mathbf{x}_i$  for all  $i$ . It follows that  $\mathbf{x}_i = A^i\mathbf{x}_0$  for all  $i$ .

The characteristic polynomial of  $A$  is

$$\begin{aligned} \det(A - \lambda I_2) &= \begin{vmatrix} 0.95 - \lambda & 0.03 \\ 0.05 & 0.97 - \lambda \end{vmatrix} \\ &= (0.95 - \lambda)(0.97 - \lambda) - 0.0015 \\ &= \lambda^2 - 1.92\lambda + 0.92. \end{aligned}$$

The solutions of the characteristic equation

$\lambda^2 - 1.92\lambda + 0.92 = 0$  are 1 and 0.92.



# Solution (cont.)

It follows that the eigenvalues of  $A$  are 1 and 0.92.

The matrix  $A - I_2 = \begin{bmatrix} -0.05 & 0.03 \\ 0.05 & -0.03 \end{bmatrix}$  is row equivalent to the matrix  $\begin{bmatrix} -5 & 3 \\ 0 & 0 \end{bmatrix}$ . It follows that  $\begin{bmatrix} 3 \\ 5 \end{bmatrix}$  is an eigenvector of  $A$  corresponding to the eigenvalue 1.

The matrix  $A - 0.92I_2 = \begin{bmatrix} 0.03 & 0.03 \\ 0.05 & 50.05 \end{bmatrix}$  is row equivalent to the matrix  $\begin{bmatrix} 1 & -1 \\ 0 & 0 \end{bmatrix}$ . It follows that  $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$  is an eigenvector of  $A$  corresponding to the eigenvalue 0.92.



# Solution (cont.)

$\mathcal{B} = \left\{ \begin{bmatrix} 3 \\ 5 \end{bmatrix}, \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right\}$  is a basis for  $\mathbb{R}^2$ , so there exists scalars  $c_1$

and  $c_2$  such that  $\mathbf{x}_0 = c_1 \begin{bmatrix} 3 \\ 5 \end{bmatrix} + c_2 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .

$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix}$  is the coordinate vector of  $\mathbf{x}_0$  with respect to  $\mathcal{B}$ , so

$$\begin{bmatrix} c_1 \\ c_2 \end{bmatrix} = \begin{bmatrix} 3 & 1 \\ 5 & -1 \end{bmatrix}^{-1} \mathbf{x}_0 = \frac{1}{-8} \begin{bmatrix} -1 & -1 \\ -5 & 3 \end{bmatrix} \begin{bmatrix} 600,000 \\ 400,000 \end{bmatrix} = \begin{bmatrix} 125,000 \\ 225,000 \end{bmatrix}$$

and  $\mathbf{x}_0 = 125,000 \begin{bmatrix} 3 \\ 5 \end{bmatrix} + 225,000 \begin{bmatrix} 1 \\ -1 \end{bmatrix}$ .



# Solution (cont.)

It follows that

$$\mathbf{x}_i = A^i \mathbf{x}_0 = A^i \left( 125,000 \begin{bmatrix} 3 \\ 5 \end{bmatrix} + 225,000 \begin{bmatrix} 1 \\ -1 \end{bmatrix} \right) =$$

$$125,000 \begin{bmatrix} 3 \\ 5 \end{bmatrix} + 225,000 \cdot (0.92)^i \begin{bmatrix} 1 \\ -1 \end{bmatrix} \text{ for all } i.$$

Since  $(0.92)^i \rightarrow 0$  as  $i \rightarrow \infty$ , it follows that

$$\mathbf{x}_i \rightarrow 125,000 \begin{bmatrix} 3 \\ 5 \end{bmatrix} = \begin{bmatrix} 375,000 \\ 625,000 \end{bmatrix} \text{ as } i \rightarrow \infty.$$





# Markov chains

- A *probability vector* is a vector with nonnegative entries that add up to 1.
- A *stochastic matrix* is a square matrix whose columns are probability vectors.
- A *Markov chain* is a sequence  $\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \dots$  of probability vectors together with a stochastic matrix  $P$  such that  $\mathbf{x}_1 = P\mathbf{x}_0, \mathbf{x}_2 = P\mathbf{x}_1, \mathbf{x}_3 = P\mathbf{x}_2, \dots$
- A *steady-state vector* (or an *equilibrium vector*) for a stochastic matrix  $P$  is a probability vector  $\mathbf{q}$  such that  $P\mathbf{q} = \mathbf{q}$ .
- So a steady-state vector for a stochastic matrix  $P$  is a probability vector which is an eigenvector of  $P$  corresponding to the eigenvalue 1.



# Problem 7 from the exam from August 2010

$$\text{Let } A = \begin{bmatrix} 2 & 0 & 2 \\ 1 & -2 & -1 \\ -1 & 6 & 5 \end{bmatrix}.$$

- 1 Solve the equation  $A\mathbf{x} = \mathbf{0}$ .
- 2 Find the eigenvalues and eigenvectors of  $A$ .



# Solution

To solve the equation  $A\mathbf{x} = \mathbf{0}$ , we reduce  $A$  to its reduced echelon form.

$$\begin{bmatrix} 2 & 0 & 2 \\ 1 & -2 & -1 \\ -1 & 6 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 1 & -2 & -1 \\ -1 & 6 & 5 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & -2 & -2 \\ 0 & 6 & 6 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 6 & 6 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$

We see that the general solution to the equation  $A\mathbf{x} = \mathbf{0}$  is

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -x_3 \\ -x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix} \text{ where } x_3 \text{ is a free parameters.}$$



# Solution (cont.)

The characteristic polynomial of  $A$  is

$$\begin{aligned}\det(A - \lambda I_3) &= \begin{vmatrix} 2 - \lambda & 0 & 2 \\ 1 & -2 - \lambda & -1 \\ -1 & 6 & 5 - \lambda \end{vmatrix} \\ &= (2 - \lambda) \begin{vmatrix} -2 - \lambda & -1 \\ 6 & 5 - \lambda \end{vmatrix} + 2 \begin{vmatrix} 1 & -2 - \lambda \\ -1 & 6 \end{vmatrix} \\ &= (2 - \lambda)((-2 - \lambda)(5 - \lambda) + 6) + 2(6 + (-2 - \lambda)) \\ &= -\lambda^3 + 5\lambda^2 - 4\lambda = -\lambda(\lambda^2 - 5\lambda + 4).\end{aligned}$$

The zeros of  $\lambda^2 - 5\lambda + 4$  are 1 and 4, so the eigenvalues of  $A$  are 0, 1 and 4.



# Solution (cont.)

It follows from the first part that the eigenvectors of  $A$  corresponding to the eigenvalue of 0 are the vectors

$$\left\{ t \begin{bmatrix} -1 \\ -1 \\ 1 \end{bmatrix} : t \neq 0 \right\}.$$



# Solution (cont.)

To find the eigenvectors of  $A$  corresponding to the eigenvalue 4, we solve the equation  $(A - 4I_3)\mathbf{x} = \mathbf{0}$ . To solve the equation  $(A - 4I_3)\mathbf{x} = \mathbf{0}$ , we reduce  $A - 4I_3$  to its reduced echelon form.

$$A - 4I_3 = \begin{bmatrix} -2 & 0 & 2 \\ 1 & -6 & -1 \\ -1 & 6 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 1 & -6 & -1 \\ -1 & 6 & 1 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & -6 & 0 \\ 0 & 6 & 0 \end{bmatrix}$$
$$\begin{bmatrix} 1 & 0 & -1 \\ 0 & -6 & 0 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -1 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$



# Solution (cont.)

We see that the solution to the equation  $(A - 4I_3)\mathbf{x} = \mathbf{0}$  is

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} x_3 \\ 0 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} \text{ where } x_3 \text{ is a free parameter. It}$$

follows that the eigenvectors of  $A$  corresponding to the

$$\text{eigenvalue of 4 are the vectors } \left\{ t \begin{bmatrix} 1 \\ 0 \\ 1 \end{bmatrix} : t \neq 0 \right\}.$$



# Solution (cont.)

To find the eigenvectors of  $A$  corresponding to the eigenvalue 1, we solve the equation  $(A - I_3)\mathbf{x} = \mathbf{0}$ . To solve the equation  $(A - I_3)\mathbf{x} = \mathbf{0}$ , we reduce  $A - I_3$  to its reduced echelon form.

$$A - I_3 = \begin{bmatrix} 1 & 0 & 2 \\ 1 & -3 & -1 \\ -1 & 6 & 4 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & -3 & -3 \\ 0 & 6 & 6 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & 2 \\ 0 & -3 & -3 \\ 0 & 0 & 0 \end{bmatrix} \rightarrow$$
$$\begin{bmatrix} 1 & 0 & 2 \\ 0 & 1 & 1 \\ 0 & 0 & 0 \end{bmatrix}$$





# Solution (cont.)

We see that the solution to the equation  $(A - I_3)\mathbf{x} = \mathbf{0}$  is

$$\mathbf{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} -2x_3 \\ -x_3 \\ x_3 \end{bmatrix} = x_3 \begin{bmatrix} -2 \\ -1 \\ 1 \end{bmatrix} \text{ where } x_3 \text{ is a free}$$

parameter. It follows that the eigenvectors of  $A$  corresponding to the eigenvalue of 1 are the vectors

$$\left\{ t \begin{bmatrix} -2 \\ -1 \\ 1 \end{bmatrix} : t \neq 0 \right\}.$$



# Problem 6 from August 2012

For which numbers  $a$  does  $\mathbb{R}^2$  have a basis of eigenvectors of the matrix  $\begin{bmatrix} 0 & a \\ 1 & 0 \end{bmatrix}$ ?



# Solution

Let  $A = \begin{bmatrix} 0 & a \\ 1 & 0 \end{bmatrix}$ . The characteristic polynomial of  $A$  is

$$\det(A - \lambda I_2) = \begin{vmatrix} -\lambda & a \\ 1 & -\lambda \end{vmatrix} = \lambda^2 - a.$$

If  $a > 0$ , then  $A$  has two distinct eigenvalues  $\pm\sqrt{a}$ . If  $\mathbf{v}_1$  is an eigenvector corresponding to  $\sqrt{a}$ , and  $\mathbf{v}_2$  is an eigenvector corresponding to  $-\sqrt{a}$ , then  $\{\mathbf{v}_1, \mathbf{v}_2\}$  is linearly independent, and therefore a basis of  $\mathbb{R}^2$ .



# Solution (cont.)

If  $a = 0$ , then  $A$  has one eigenvalue 0. The corresponding eigenspace is  $\text{Nul}(A) = \text{Span} \left\{ \begin{bmatrix} 0 \\ 1 \end{bmatrix} \right\}$  which is one-dimensional. It follows that all eigenvectors of  $A$  are linearly dependent, and thus that  $\mathbb{R}^2$  does not have a basis of eigenvectors of  $A$  in this case.

If  $a < 0$ , then  $A$  does not have any real eigenvalues, and therefore  $\mathbb{R}^2$  does not have a basis of eigenvectors of  $A$  in this case ( $A$  has complex eigenvalues, but the corresponding eigenvectors will not belong to  $\mathbb{R}^2$  and therefore no eigenvectors of  $A$  will form a basis for  $\mathbb{R}^2$ ).

So  $\mathbb{R}^2$  has a basis of eigenvectors of  $A$  if and only if  $a > 0$ .



# Tomorrow's lecture

Tomorrow we shall look at

- *diagonal* matrices,
- how to *diagonalizable* a matrix.

Section 5.3 in “Linear Algebras and Its Applications” (pages 281—288).

