



# DAMA 50

## Written Assignment III

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# [DAMA 50] Written Assignment 3

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## Problem 7

Consider the matrix

$$A = \begin{pmatrix} -2 & 6 & -3 \\ 6 & 3 & 2 \\ -3 & 2 & a \end{pmatrix} \quad (1.1)$$

where  $a$  is a real parameter and  $x = (1, 1, 1)^T$ .

Consider the case  $a = -18$ . Using **Sagemath**

- (a) compute the eigenvalues and the eigenvectors of  $A$ ;
- (b) normalize the eigenvectors of  $A$ ;
- (c) compute (numerically with 3 significant digits) the components of  $x$  in the orthonormal basis of question (b).

Consider the case  $a = 6$ . Using **Sagemath**

- (d) compute the eigenvalues and the eigenvectors of  $A$ ;
  - (e) what is the algebraic multiplicity of each eigenvalue?
  - (f) are the eigenvectors orthogonal? If not find an orthonormal basis.
  - (g) compute (numerically with 3 significant digits) the components of  $x$  in the orthonormal basis of question (f).
- 

```
In [1]: restore()
```

```
In [2]: %display latex
```

For  $a = -18$

```
In [3]: A=matrix([[-2,6,-3],[6,3,2],[-3,2,-18]]);A
```

```
Out[3]: 
$$\begin{pmatrix} -2 & 6 & -3 \\ 6 & 3 & 2 \\ -3 & 2 & -18 \end{pmatrix}$$

```

```
In [4]: x=vector([1,1,1]);x.column()
```

```
Out[4]:  $\begin{pmatrix} 1 \\ 1 \\ 1 \end{pmatrix}$ 
```

(a)

Compute the diagonal matrix of eigenvalues  $eA$  which bears all the eigenvalues in its diagonal and the right eigenmatrix  $vA$  with all the eigenvectors as its columns

```
In [5]: eA,vA=A.eigenmatrix_right();
```

```
In [6]: eA
```

```
Out[6]:  $\begin{pmatrix} 7 & 0 & 0 \\ 0 & -5 & 0 \\ 0 & 0 & -19 \end{pmatrix}$ 
```

```
In [7]: vA
```

```
Out[7]:  $\begin{pmatrix} 1 & 1 & 1 \\ \frac{3}{2} & -\frac{2}{3} & -\frac{2}{3} \\ 0 & -\frac{1}{3} & \frac{13}{3} \end{pmatrix}$ 
```

Extract the eigenvalues

```
In [8]: Aeigenvalues=A.eigenvalues();Aeigenvalues
```

```
Out[8]:  $[7, -5, -19]$ 
```

Matrix A eigenvalues are :  $\lambda_1 = 7, \lambda_2 = -5, \lambda_3 = -19$

Extract the eigenvectors

```
In [9]: A.eigenvectors_right()
```

```
Out[9]:  $\left[ \left( 7, \left[ \left( 1, \frac{3}{2}, 0 \right) \right], 1 \right), \left( -5, \left[ \left( 1, -\frac{2}{3}, -\frac{1}{3} \right) \right], 1 \right), \left( -19, \left[ \left( 1, -\frac{2}{3}, \frac{13}{3} \right) \right], 1 \right) \right]$ 
```

```
In [10]: Ae1=vector(vA.column(0))
Ae2=vector(vA.column(1))
Ae3=vector(vA.column(2));Ae1,Ae2,Ae3
```

```
Out[10]:  $\left( \left( 1, \frac{3}{2}, 0 \right), \left( 1, -\frac{2}{3}, -\frac{1}{3} \right), \left( 1, -\frac{2}{3}, \frac{13}{3} \right) \right)$ 
```

Matrix A eigenvectors are :  $\vec{v}_1 = [1, \frac{3}{2}, 0]^T$      $\vec{v}_2 = [1, -\frac{2}{3}, -\frac{1}{3}]^T$      $\vec{v}_3 = [1, -\frac{2}{3}, \frac{13}{3}]^T$

(b)

Normalize the eigenvectors

```
In [11]: An1=Ae1.normalized()
An2=Ae2.normalized()
An3=Ae3.normalized();An1,An2,An3
```

```
Out[11]:  $\left( \left( \frac{2}{13} \sqrt{13}, \frac{3}{13} \sqrt{13}, 0 \right), \left( \frac{3}{14} \sqrt{14}, -\frac{1}{7} \sqrt{14}, -\frac{1}{14} \sqrt{14} \right), \left( \frac{3}{182} \sqrt{182}, -\frac{1}{91} \sqrt{182}, \frac{1}{14} \sqrt{182} \right) \right)$ 
```

The vectors An1, An2, An3 form an orthonormal basis

**(c)**

To compute the components of  $x$  in the ONB mentioned above, we take its dot product with each vector of the ONB

```
In [12]: x_comp = [x.dot_product(An1), x.dot_product(An2), x.dot_product(An3)];x_comp
```

```
Out[12]:  $\left[ \frac{5}{13} \sqrt{13}, 0, \frac{1}{13} \sqrt{182} \right]$ 
```

```
In [13]: x_comp_num = [x_comp[0].n(digits=3), x_comp[1].n(digits=3), x_comp[2].n(digits=3)];x_comp_num
```

```
Out[13]: [1.39, 0.000, 1.04]
```

**(d)**

For  $a = 6$

```
In [14]: A2=matrix([[-2,6,-3],[6,3,2],[-3,2,6]]);A2
```

```
Out[14]:  $\begin{pmatrix} -2 & 6 & -3 \\ 6 & 3 & 2 \\ -3 & 2 & 6 \end{pmatrix}$ 
```

Compute the diagonal matrix of eigenvalues  $eA2$  which bears all the eigenvalues in its diagonal and the right eigenmatrix  $vA2$  with all the eigenvectors as its columns

```
In [15]: eA2,vA2=A2.eigenmatrix_right()
```

```
In [16]: eA2
```

```
Out[16]:  $\begin{pmatrix} -7 & 0 & 0 \\ 0 & 7 & 0 \\ 0 & 0 & 7 \end{pmatrix}$ 
```

```
In [17]: vA2
```

```
Out[17]:  $\begin{pmatrix} 1 & 1 & 0 \\ -\frac{2}{3} & 0 & 1 \\ \frac{1}{3} & -3 & 2 \end{pmatrix}$ 
```

Extract the eigenvalues

```
In [18]: A2eigenvalues=A2.eigenvalues();A2eigenvalues
```

```
Out[18]: [-7, 7, 7]
```

Sort the eigenvalues in descending order

```
In [19]: A2eigenvalues_ord = [A2eigenvalues[1],A2eigenvalues[2],A2eigenvalues[0]];A2eigenvalues_ord
Out[19]: [7, 7, -7]
```

Matrix A2 eigenvalues are :  $\lambda_1 = 7$  with algebraic multiplicity 2 and  $\lambda_2 = -7$

Extract the eigenvectors

```
In [20]: A2eigen = A2.eigenvectors_right();A2eigen
Out[20]:  $\left[ \left( -7, \left[ \left( 1, -\frac{2}{3}, \frac{1}{3} \right) \right], 1 \right), (7, [(1, 0, -3), (0, 1, 2)], 2) \right]$ 
```

Store the eigenvectors in descending eigenvalue order

```
In [21]: A2e1=vector(vA2.column(1))
A2e2=vector(vA2.column(2))
A2e3=vector(vA2.column(0));A2e1,A2e2,A2e3
Out[21]:  $\left( (1, 0, -3), (0, 1, 2), \left( 1, -\frac{2}{3}, \frac{1}{3} \right) \right)$ 
```

Matrix A eigenvectors are :  $\vec{v}_1 = [1, 0, -3]^T$      $\vec{v}_2 = [0, 1, 2]^T$      $\vec{v}_3 = [1, -\frac{2}{3}, \frac{1}{3}]^T$

(e)

```
In [22]: print(f"Eigenvalue {A2eigen[0][0]} has an algebraic multiplicity of {A2eigen[0][2]}")
print(f"Eigenvalue {A2eigen[1][0]} has an algebraic multiplicity of {A2eigen[1][2]}")

Eigenvalue -7 has an algebraic multiplicity of 1
Eigenvalue 7 has an algebraic multiplicity of 2
```

(f)

Checking if all dot products among the eigenvectors are equal to 0

```
In [23]: print(f"The dot products are {A2e1*A2e2},{A2e1*A2e3},{A2e2*A2e3}")

The dot products are -6,0,0
```

We notice that the eigenvectors  $\vec{v}_1 = [1, 0, -3]^T$  and  $\vec{v}_2 = [0, 1, 2]^T$  have a dot product of  $-6$  which means that they are not orthogonal

Compute an orthogonal basis using the Gram-Schmidt orthogonalization process

```
In [24]: u1 = A2e1
u2 = A2e2 - ((A2e2.dot_product(u1)) / (u1.dot_product(u1))) * u1
u3 = A2e3 - ((A2e3.dot_product(u1)) / (u1.dot_product(u1))) * u1 - ((A2e3.dot_product(u2)) / (u2.d
ot_product(u2))) * u2;u1,u2,u3
Out[24]:  $\left( (1, 0, -3), \left( \frac{3}{5}, 1, \frac{1}{5} \right), \left( 1, -\frac{2}{3}, \frac{1}{3} \right) \right)$ 
```

Normalize the vectors of the orthogonal basis to get an orthonormal basis

```
In [25]: A2_orthonormal_basis=[u1.normalized(),u2.normalized(),u3.normalized()];A2_orthonormal_basis
```

```
Out[25]:  $\left[ \left( \frac{1}{10} \sqrt{10}, 0, -\frac{3}{10} \sqrt{10} \right), \left( \frac{3}{7} \sqrt{\frac{7}{5}}, \frac{5}{7} \sqrt{\frac{7}{5}}, \frac{1}{7} \sqrt{\frac{7}{5}} \right), \left( \frac{3}{14} \sqrt{14}, -\frac{1}{7} \sqrt{14}, \frac{1}{14} \sqrt{14} \right) \right]$ 
```

(g)

To compute the components of  $x$  in the ONB mentioned above, we take its dot product with each vector of the ONB

```
In [26]: x_comp2 = [x.dot_product(u1.normalized()), x.dot_product(u2.normalized()), x.dot_product(u3.normalized())];x_comp2
```

```
Out[26]:  $\left[ -\frac{1}{5} \sqrt{10}, \frac{9}{7} \sqrt{\frac{7}{5}}, \frac{1}{7} \sqrt{14} \right]$ 
```

```
In [27]: x_comp2_num = [x_comp2[0].n(digits=3), x_comp2[1].n(digits=3), x_comp2[2].n(digits=3)];x_comp2_num
```

```
Out[27]:  $[-0.632, 1.52, 0.534]$ 
```

## [DAMA 50] Written Assignment 3

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### Problem 8

Consider the  $6 \times 2$  matrix  $A$  with elements

$$a_{ij} = i^2 + 2ij, \quad i = 1, \dots, 6, \quad j = 1, 2$$

Using `sagemath`

- Compute  $A$ ;
  - Compute the SVD of  $A$  and print the left and right singular vector matrices and the singular value matrix using 3 significant digits;
  - Compute the rank-1 approximation  $A_1$  of  $A$  and print using 3 significant digits;
  - Compute the Frobenius norm of the difference  $A - A_1$  (e) Compute the rank-2 approximation  $A_2$  of  $A$  and print using 3 significant digits. Compute the Frobenius norm of the difference  $A - A_2$ . Comment on these results.
-

## Solution

```
In [28]: restore()
```

```
In [29]: %display latex
```

(a)

Compute matrix  $A$

```
In [30]: A=matrix(QQ,6,2, lambda i,j: ((i+1)^2+2*(i+1)*(j+1)));A
```

```
Out[30]: 
$$\begin{pmatrix} 3 & 5 \\ 8 & 12 \\ 15 & 21 \\ 24 & 32 \\ 35 & 45 \\ 48 & 60 \end{pmatrix}$$

```

(b)

Compute the left singular vector matrix  $U$ , the singular value matrix  $S$  and the right singular vector matrix  $V$  for the SVD of  $A$

```
In [31]: U,S,V = (A.change_ring(RDF)).SVD()
```

```
In [32]: U.n(digits=3)
```

```
Out[32]: 
$$\begin{pmatrix} -0.0536 & -0.340 & -0.293 & -0.398 & -0.506 & -0.618 \\ -0.133 & -0.511 & -0.480 & -0.286 & 0.0854 & 0.634 \\ -0.239 & -0.514 & 0.802 & -0.159 & -0.0659 & 0.0796 \\ -0.370 & -0.349 & -0.174 & 0.823 & -0.152 & -0.0992 \\ -0.528 & -0.0152 & -0.104 & -0.177 & 0.731 & -0.381 \\ -0.712 & 0.487 & 0.0107 & -0.160 & -0.419 & 0.235 \end{pmatrix}$$

```

```
In [33]: S.n(digits=3)
```

```
Out[33]: 
$$\begin{pmatrix} 108. & 0.000 \\ 0.000 & 2.07 \\ 0.000 & 0.000 \\ 0.000 & 0.000 \\ 0.000 & 0.000 \\ 0.000 & 0.000 \end{pmatrix}$$

```

```
In [34]: V.n(digits=3)
```

```
Out[34]: 
$$\begin{pmatrix} -0.614 & 0.789 \\ -0.789 & -0.614 \end{pmatrix}$$

```

(c)

We construct the matrix  $A_1$  as  $A_1 = u_1 v_1^T$  which is formed by the outer product of the first orthogonal column vector of  $U$  and  $V$

```
In [35]: u1=U[:,0];u1.n(digits=3)
```

```
Out[35]: 
$$\begin{pmatrix} -0.0536 \\ -0.133 \\ -0.239 \\ -0.370 \\ -0.528 \\ -0.712 \end{pmatrix}$$

```

```
In [36]: v1T=V[:,0].transpose();v1T.n(digits=3)
```

```
Out[36]:  $(-0.614 \quad -0.789)$ 
```

```
In [37]: A1=u1*v1T;A1.n(digits=3)
```

```
Out[37]: 
$$\begin{pmatrix} 0.0329 & 0.0423 \\ 0.0818 & 0.105 \\ 0.147 & 0.188 \\ 0.228 & 0.292 \\ 0.324 & 0.417 \\ 0.437 & 0.561 \end{pmatrix}$$

```

We compute the rank-1 approximation  $\hat{A}(1) := \sigma_1 A_1$  where  $\sigma_1$  is the first singular value

```
In [38]:  $\sigma_1=S[0][0];\sigma_1.n(digits=3)$ 
```

```
Out[38]: 108.
```

```
In [39]: A1aprx= $\sigma_1$ *A1;A1aprx.n(digits=3)
```

```
Out[39]: 
$$\begin{pmatrix} 3.56 & 4.57 \\ 8.84 & 11.3 \\ 15.8 & 20.3 \\ 24.6 & 31.6 \\ 35.0 & 45.0 \\ 47.2 & 60.6 \end{pmatrix}$$

```

(d)

Compute the difference  $A - A_1$  and store it in a matrix  $B1$

```
In [40]: B1=A-A1aprx;B1.n(digits=3)
```

```
Out[40]: 
$$\begin{pmatrix} -0.556 & 0.433 \\ -0.837 & 0.651 \\ -0.842 & 0.655 \\ -0.571 & 0.445 \\ -0.0249 & 0.0194 \\ 0.797 & -0.620 \end{pmatrix}$$

```

Compute the Frobenius norm of  $B1$

```
In [41]: frobnorm1 = B1.norm('frob');frobnorm1
```

```
Out[41]: 2.0746347947807724
```



```
In [42]: frobnorm1.n(digits=3)
```

```
Out[42]: 2.07
```

(e)

For the rank-2 approximation we will need to compute  $\hat{A}(2) := \sigma_1 A_1 + \sigma_2 A_2$

```
In [43]: u2=U[:,1];u2.n(digits=3)
```

```
Out[43]: 
$$\begin{pmatrix} -0.340 \\ -0.511 \\ -0.514 \\ -0.349 \\ -0.0152 \\ 0.487 \end{pmatrix}$$

```

```
In [44]: v2T=V[:,1].transpose();v2T.n(digits=3)
```

```
Out[44]:  $(0.789 \quad -0.614)$ 
```

```
In [45]: A2=u2*v2T;A2.n(digits=3)
```

```
Out[45]: 
$$\begin{pmatrix} -0.268 & 0.209 \\ -0.403 & 0.314 \\ -0.406 & 0.316 \\ -0.275 & 0.214 \\ -0.0120 & 0.00936 \\ 0.384 & -0.299 \end{pmatrix}$$

```

```
In [46]:  $\sigma_2=S[1][1];\sigma_2.n(digits=3)$ 
```

```
Out[46]: 2.07
```

$\hat{A}(2) := \sigma_1 A_1 + \sigma_2 A_2$

```
In [47]: A2aprx =  $\sigma_1 A_1 + \sigma_2 A_2$ ;A2aprx.n(digits=3)
```

```
Out[47]: 
$$\begin{pmatrix} 3.00 & 5.00 \\ 8.00 & 12.0 \\ 15.0 & 21.0 \\ 24.0 & 32.0 \\ 35.0 & 45.0 \\ 48.0 & 60.0 \end{pmatrix}$$

```

Compute the difference  $A - A_2$  and store it in a matrix  $B2$

```
In [48]: B2=A-A2aprx;B2.n(digits=3)
```

```
Out[48]: 
$$\begin{pmatrix} 2.66 \times 10^{-15} & 4.44 \times 10^{-15} \\ 2.66 \times 10^{-15} & 3.55 \times 10^{-15} \\ 3.55 \times 10^{-15} & 3.55 \times 10^{-15} \\ 3.55 \times 10^{-15} & 3.55 \times 10^{-15} \\ 7.11 \times 10^{-15} & 7.11 \times 10^{-15} \\ 7.11 \times 10^{-15} & 7.11 \times 10^{-15} \end{pmatrix}$$

```

Compute the Frobenius norm of  $B2$

```
In [49]: frobnorm2 = B2.norm('frob');frobnorm2
```

```
Out[49]: 1.7290988493396307 × 10-14
```

```
In [50]: frobnorm2.n(digits=3)
```

```
Out[50]: 1.73 × 10-14
```

## Comment

The matrices  $B_1 = A - A_1$  and  $B_2 = A - A_2$  capture the information that cannot be represented by the rank-1 and rank-2 approximation respectively (the error). The Frobenius norm of a matrix  $A$ , which is defined by  $\|A\|_F = \left( \sum_{i=1}^n \sum_{j=1}^n a_{i,j}^2 \right)^{\frac{1}{2}}$  can give us an idea of the magnitude of the errors in the matrices. We observe that the error that corresponds to the rank-2 approximation is significantly smaller than the error that corresponds to the rank-1 approximation, which means that the rank-2 captures the essential features of the original matrix better than the rank-1 approximation.

## [DAMA 50] Written Assignment 3

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### Problem 9

Given the eigenvalues  $\lambda_1 = 9, \lambda_2 = 0, \lambda_3 = -9$  and the associated eigenvectors  $e_1 = \frac{1}{3}(-2, 2, 1)$ ,  $e_2 = \frac{1}{3}(2, 1, 2)$ ,  $e_3 = \frac{1}{3}(-1, -2, 2)$  of a real symmetric matrix  $A$ , compute  $A$  using by-hand calculation.

---

**Theorem 4.20** (Eigendecomposition). *A square matrix  $A \in \mathbb{R}^{n \times n}$  can be factored into*

$$A = PDP^{-1}, \quad (4.55)$$

*where  $P \in \mathbb{R}^{n \times n}$  and  $D$  is a diagonal matrix whose diagonal entries are the eigenvalues of  $A$ , if and only if the eigenvectors of  $A$  form a basis of  $\mathbb{R}^n$ .*

**Theorem 4.21.** *A symmetric matrix  $S \in \mathbb{R}^{n \times n}$  can always be diagonalized.*

Since we have a real symmetric matrix  $A$ , it will be a square matrix and given its eigenvalues and eigenvectors we can compute the matrices  $P$  and  $D$  mentioned in Theorem (4.20) such that:

$$P = \begin{pmatrix} -2/3 & 2/3 & -1/3 \\ 2/3 & 1/3 & -2/3 \\ 1/3 & 2/3 & 2/3 \end{pmatrix} = \frac{1}{3} \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix} \quad \text{and} \quad D = \begin{pmatrix} 9 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -9 \end{pmatrix} = 9 \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix}$$

Therefore, we can compute  $A$  which will be:

$$\begin{aligned} A &= P \cdot D \cdot P^T = \frac{1}{3} \cdot \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix} \cdot 9 \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \frac{1}{3} \cdot \begin{pmatrix} -2 & 2 & 1 \\ 2 & 1 & 2 \\ -1 & -2 & 2 \end{pmatrix} = \\ &= \frac{9}{9} \cdot \begin{pmatrix} -2 & 2 & -1 \\ 2 & 1 & -2 \\ 1 & 2 & 2 \end{pmatrix} \cdot \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 0 \\ 0 & 0 & -1 \end{pmatrix} \cdot \begin{pmatrix} -2 & 2 & 1 \\ 2 & 1 & 2 \\ -1 & -2 & 2 \end{pmatrix} = \begin{pmatrix} -2 & 0 & 1 \\ 2 & 0 & 2 \\ 1 & 0 & -2 \end{pmatrix} \cdot \begin{pmatrix} -2 & 2 & 1 \\ 2 & 1 & 2 \\ -1 & -2 & 2 \end{pmatrix} = \\ &= \begin{pmatrix} 3 & -6 & 0 \\ -6 & 0 & 6 \\ 0 & 6 & -3 \end{pmatrix} \end{aligned}$$

## [DAMA 50] Written Assignment 3

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### Problem 10

Given the matrix  $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ , answer the following questions, by providing all the necessary by-hand calculations:

- (i) Compute the characteristic polynomial and the eigenvalues of the matrix  $A$ .
  - (ii) Compute the eigenvectors of the matrix  $A$ .
  - (iii) Find a general formula for the matrix  $A^n$  with  $n \in \mathbb{Z}^+$ , utilizing the diagonalizable nature of the original matrix  $A$ .
- 

(i) We are given matrix  $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$

The characteristic polynomial will be  $p_A(\lambda) := \det(A - \lambda I)$  where

$$\det(A - \lambda I) = \begin{vmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{vmatrix} = (2-\lambda)^2 - 1 = \lambda^2 - 4\lambda + 4 - 1 = \lambda^2 - 4\lambda + 3$$

$$p_A(\lambda) = \lambda^2 - 4\lambda + 3$$

The eigenvalues of  $A$  are the roots of the characteristic polynomial.

$$p_A(\lambda) = 0 \Leftrightarrow \lambda^2 - 4\lambda + 3 = 0 \Leftrightarrow (\lambda - 3)(\lambda - 1) = 0 \Leftrightarrow \lambda = 3 \text{ or } \lambda = 1$$

- (ii) We find the eigenvectors that correspond to these eigenvalues by looking at vectors  $\mathbf{x}$  such that:

$$\begin{pmatrix} 2-\lambda & 1 \\ 1 & 2-\lambda \end{pmatrix} \mathbf{x} = 0$$

For  $\lambda=3$  we will solve the homogenous system:

$$\begin{pmatrix} 2-3 & 1 \\ 1 & 2-3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \Leftrightarrow \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \Leftrightarrow \begin{cases} -x_1 + x_2 = 0 \\ x_1 - x_2 = 0 \end{cases} \Leftrightarrow$$

$$\Leftrightarrow \begin{cases} x_1 = x_2 \\ x_1 = x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = x_2 = t \\ x_2 = t \end{cases} \Leftrightarrow \vec{x} = \begin{pmatrix} 1 \\ 1 \end{pmatrix} t$$

so we obtain a solution space  $E_3 = \text{span} \left\{ \begin{pmatrix} 1 \\ 1 \end{pmatrix} \right\}$

For  $\lambda=1$  we will solve the homogenous system:

$$\begin{pmatrix} 2 & -1 \\ 1 & 2 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \Leftrightarrow \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = 0 \Leftrightarrow \begin{cases} x_1 + x_2 = 0 \\ x_1 + x_2 = 0 \end{cases} \Leftrightarrow$$

$$\Leftrightarrow \begin{cases} x_1 = -x_2 \\ x_1 = -x_2 \end{cases} \Leftrightarrow \begin{cases} x_1 = -x_2 = -t \\ x_2 = t \end{cases} \Leftrightarrow \vec{x} = \begin{pmatrix} -1 \\ 1 \end{pmatrix} t$$

so we obtain a solution space  $E_1 = \text{span} \left\{ \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}$

Therefore, the eigenvalues of  $A$  are  $\lambda_1 = 3$  and  $\lambda_2 = 1$  (the roots of the characteristic polynomial), and the associated eigenvectors

are  $\vec{e}_1 = \begin{pmatrix} 1 \\ 1 \end{pmatrix}$  and  $\vec{e}_2 = \begin{pmatrix} -1 \\ 1 \end{pmatrix}$

(iii)

**Theorem 4.20** (Eigendecomposition). *A square matrix  $A \in \mathbb{R}^{n \times n}$  can be factored into*

$$A = PDP^{-1}, \quad (4.55)$$

where  $P \in \mathbb{R}^{n \times n}$  and  $D$  is a diagonal matrix whose diagonal entries are the eigenvalues of  $A$ , if and only if the eigenvectors of  $A$  form a basis of  $\mathbb{R}^n$ .

$$\det(\vec{e}_1, \vec{e}_2) = \begin{vmatrix} 1 & -1 \\ 1 & 1 \end{vmatrix} = 1 + 1 = 2 \neq 0. \text{ So } \vec{e}_1 \text{ and } \vec{e}_2 \text{ are linearly independent so}$$

they form a basis of  $\mathbb{R}^2$ . Therefore, according to theorem (4.55), the matrix  $A$  can be factored into  $A = PDP^{-1}$

$$\text{where } P = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix}, D = \begin{pmatrix} 3 & 0 \\ 0 & 1 \end{pmatrix} \text{ and } P^{-1} = \begin{pmatrix} 1/2 & 1/2 \\ -1/2 & 1/2 \end{pmatrix} = \frac{1}{2} \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} \text{ since}$$

$$\begin{pmatrix} 1 & -1 & | & 1 & 0 \\ 1 & 1 & | & 0 & 1 \end{pmatrix} \xrightarrow{R2 \leftarrow R2 - R1} \begin{pmatrix} 1 & -1 & | & 1 & 0 \\ 0 & 2 & | & -1 & 1 \end{pmatrix} \xrightarrow{R2 \leftarrow \frac{R2}{2}} \begin{pmatrix} 1 & -1 & | & 1 & 0 \\ 0 & 1 & | & -1/2 & 1/2 \end{pmatrix} =$$

$$\xrightarrow{R1 \leftarrow R1 + R2} \begin{pmatrix} 1 & 0 & | & 1/2 & 1/2 \\ 0 & 1 & | & -1/2 & 1/2 \end{pmatrix}$$

Also according to the property (4.62) we can compute  $A^n$

- Diagonal matrices  $D$  can efficiently be raised to a power. Therefore, we can find a matrix power for a matrix  $A \in \mathbb{R}^{n \times n}$  via the eigenvalue decomposition (if it exists) so that

$$A^k = (PDP^{-1})^k = PD^kP^{-1}. \quad (4.62)$$

Computing  $D^k$  is efficient because we apply this operation individually to any diagonal element.

$$A^n = P \cdot D^n \cdot P^{-1} = \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \cdot \begin{pmatrix} 3^n & 0 \\ 0 & 1^n \end{pmatrix} \cdot \frac{1}{2} \cdot \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} = \frac{1}{2} \cdot \begin{pmatrix} 3^n & -1 \\ 3^n & 1 \end{pmatrix} \cdot \begin{pmatrix} 1 & 1 \\ -1 & 1 \end{pmatrix} = \frac{1}{2} \cdot \begin{pmatrix} 3^n + 1 & 3^n - 1 \\ 3^n - 1 & 3^n + 1 \end{pmatrix}$$