

# Proseminar on computer-assisted mathematics

## Session 4 – Kernels, images and diagonalisation in Sagemath

```
# Example 1
A = matrix(QQ, [[2,0,4],[3,-4,12],[1,-2,5]])
f_A = A.charpoly("t")
show( f_A )

 $t^3 - 3t^2 + 2t$ 

# We can factorise f_A
show( f_A.factor() )

 $(t - 2) \cdot (t - 1) \cdot t$ 

# And its roots are indeed the eigenvalues of A
ev_A = A.eigenvalues()
show( ev_A )

[2, 1, 0]
```

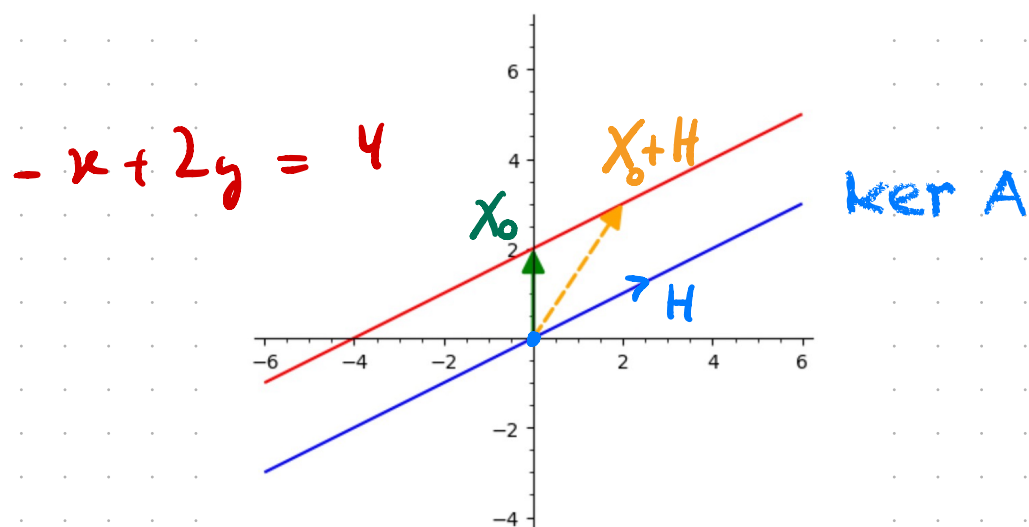
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What we want to be able to do:

- **Parameterise** the set of solutions of a non-homogeneous linear system  $AX = Y$  (which is an affine space).
- **Extract**, from a family of vectors, a basis of the subspace that they generate.
- **Complete** a basis of a subspace to a basis of the ambient space.
- **Determine** whether a given matrix is diagonalisable and, if so, **construct** a basis of eigenvectors and the associated eigenvalues.

# 1. Kernels and images

Recall that the set of all solutions of a linear system  $AX = Y$  is an affine space of direction  $\ker A$ .



$$A = \begin{pmatrix} -1 & 2 \end{pmatrix} \quad Y = 4$$

$$X_0 = \begin{pmatrix} 0 \\ 2 \end{pmatrix} \quad \ker A = \text{span}_{\mathbb{R}} \left( \begin{bmatrix} 2 \\ 1 \end{bmatrix} \right)$$

Proof:

Assume  $\exists X_0, AX_0 = Y$ .

Then, for all  $X$ ,

$$AX = Y \text{ iff } A(X - X_0) = 0$$

$$\text{i.e. } H := (X - X_0) \in \ker A.$$

$$\text{So } AX = Y$$

iff

$$\exists H \in \ker A, X = X_0 + H. \quad \square$$

$$AX = Y$$

iff

$$\exists t \in \mathbb{R}, X = \begin{bmatrix} 0 \\ 2 \end{bmatrix} + t \begin{bmatrix} 2 \\ 1 \end{bmatrix}$$

So, to solve  $AX = Y$ , we need to find one particular solution of that equation, as well as the general solution of the equation  $AX = 0$ .

Example :

$$\underbrace{\begin{pmatrix} 1 & 1 & -1 & 5 \\ 0 & -1 & 3 & 0 \end{pmatrix}}_A \underbrace{\begin{pmatrix} x \\ y \\ z \\ t \end{pmatrix}}_Y = \underbrace{\begin{pmatrix} 2 \\ -1 \end{pmatrix}}_Y$$

Both can be obtained from the **Gaussian reduction** of the augmented matrix  $(A|Y)$ .

```
y = vector( QQ, [ 2, -1 ] )  
M = A.augment( y, subdivide = True )  
show( M )
```

$$\left( \begin{array}{cccc|c} 1 & 1 & -1 & 5 & 2 \\ 0 & -1 & 3 & 0 & -1 \end{array} \right)$$

```
show( M.echelon_form() )
```

$$\left( \begin{array}{cccc|c} 1 & 0 & 2 & 5 & 1 \\ 0 & 1 & -3 & 0 & 1 \end{array} \right)$$

$\rightarrow X_0 := \begin{bmatrix} 1 \\ 1 \\ 0 \\ 0 \end{bmatrix}$  satisfies  $AX_0 = Y$

$\rightarrow \ker A = \text{span} \left( \begin{bmatrix} -2 \\ 3 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -5 \\ 0 \\ 0 \\ 1 \end{bmatrix} \right)$

The Gaussian reduction can also be used to:

- Find a basis of the column space of a matrix.
- Find linear dependence relations between the columns of a matrix.
- Complete a family of linearly independent vectors to a basis of the ambient space.

```
# Let us retake the previous matrix A
show(A)
```

$$\begin{pmatrix} 1 & 1 & -1 & 5 \\ 0 & -1 & 3 & 0 \end{pmatrix}$$

```
# The rank of A is equal to the number of the number of pivots in A1
A1 = A.echelon_form()
show(A1)
```

$$\begin{pmatrix} 1 & 0 & 2 & 5 \\ 0 & 1 & -3 & 0 \end{pmatrix}$$

$(C_1(A), C_2(A))$  is a basis  
of the column space.

Moreover :

$$C_3(A) = 2C_1(A) - 3C_2(A)$$

$$C_4(A) = 5C_1(A) + 0C_2(A)$$

## 2. Diagonalisation

**Definition.** Let  $\mathbb{k}$  be a field and let  $n > 0$  be an integer. A matrix  $A \in \text{Mat}(n \times n; \mathbb{k})$  is called **diagonalisable over  $\mathbb{k}$**  if there exists a pair of matrices  $(D, P)$  in  $\text{Mat}(n \times n; \mathbb{k})$  such that:

1.  $D$  is diagonal.
2.  $P$  is invertible.
3.  $AP = PD$ .

The last equality means that, for all  $j \in \{1; \dots; n\}$ , the  $j$ -th column of  $P$  is an eigenvector for  $A$ , associated to the  $j$ -th diagonal coefficient  $d_j$  of  $D$ :

$$\forall j \in \{1; \dots; n\}, AC_j(P) = d_j C_j(P)$$

where

$$D = \begin{pmatrix} d_1 & & \\ & \ddots & \\ & & d_n \end{pmatrix}$$

and  $P = [C_1(P), \dots, C_n(P)]$ .

$$A \begin{pmatrix} | & & | \\ C_1(P) & \dots & C_n(P) \\ | & & | \end{pmatrix} = \begin{pmatrix} | & & | \\ C_1(P) & \dots & C_n(P) \\ | & & | \end{pmatrix} \begin{pmatrix} d_1 & & \\ & \ddots & \\ & & d_n \end{pmatrix}$$
$$= \begin{pmatrix} | & & | \\ d_1 C_1(P) & \dots & d_n C_n(P) \\ | & & | \end{pmatrix}$$

**Theorem** A matrix  $A \in \text{Mat}(n \times n; \mathbb{K})$  is diagonalisable over  $\mathbb{K}$  if and only if its characteristic polynomial

$$f_A(t) := \det(tI_n - A)$$

splits into a product of linear factors

$$f_A(t) = (t - a_1)^{m_1} \dots (t - a_r)^{m_r}, \quad a_j \in \mathbb{K}$$

and

$$\forall j \in \{1; \dots; r\}, \quad \dim \ker(A - a_j I_n) = m_j.$$

In other words,  $A$  is diagonalisable over  $\mathbb{K}$  if and only if its characteristic polynomial  $f_A(t)$  splits over  $\mathbb{K}$  and the geometric multiplicity of  $a_j$  as an eigenvalue of  $A$  is equal to its algebraic multiplicity as a root of  $f_A(t)$ .

We will now see how to apply this theorem using Sage. Note that sometimes the characteristic polynomial of  $A$  is defined as  $\det(A - tI_n)$ , which is equal to  $(-1)^n \times f_A(t)$  with  $f_A(t)$  as above. We have chosen to follow Sage's convention here.

```
# Example 2, with multiple eigenvalues
A = matrix(QQ, [[2,-3,1],[1,-2,1],[1,-3,2]])
f_A = A.charpoly("t")
show( f_A.factor() )
```

$$t \cdot (t - 1)^2$$

```
# Sage can show us the eigenvalues of A, counted with their respective multiplicities
show( A.eigenvalues() )
```

$$[0, 1, 1]$$

```
# Similarly, it can show us eigenvectors for A
show( A.eigenvectors_right() )
```

$$[(0, [(1, 1, 1)], 1), (1, [(1, 0, -1), (0, 1, 3)], 2)]$$

The eigenvalue
   
a basis of the corresponding eigenspace
   
The (algebraic) multiplicity

```
D, P = A.eigenmatrix_right()
show( D, P )
```

$$\begin{pmatrix} 0 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & -1 & 3 \end{pmatrix}$$

To check :  $AP = PD$