

UNIVERSITÉ DE GENÈVE

DATA SCIENCE

TP 1: Linear Algebra

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1 - Matrix

.1

When the number of equations, here 3, is strictly larger than the number of variables, here 2, the equations system has no solution.

.2

$$\begin{cases} x + ay = 2 \\ bx + 2y = 3 \end{cases} \Leftrightarrow \begin{cases} x = 2 - ay \\ bx + 2y = 3 \end{cases}$$

We can now insert the 1st equation into the 2nd equation as follows:

$$b(2 - ay) + 2y = 3 \Leftrightarrow 2b - aby + 2y = 3 \Leftrightarrow 2b - y(ab - 2) = 3 \Leftrightarrow y = (-1) * \left(\frac{3-2b}{(ab-2)} \right)$$

With given a and b values we can easily compute y , then we reinsert y into the first equation to get x .

2 - The importance of the mathematical concept behind a code

.1

`def project_on_first(u, v)` receives two column vectors as an argument, and it projects v onto u , the projected vector is usually called v' . Visually, it means that v' and u are collinear. This also means that: $\exists \alpha$ tq. $v' * \alpha = u$.

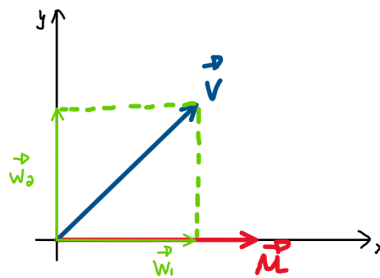


Figure 1: Projection of \vec{v} onto $\vec{u} = \vec{w}_1$

.2

`zip()` function takes as argument two python lists of same size. It then merges one value from the first list, with another value from the second list (same index), creating a list of tuples.

Let's see an example:

$$x = \text{zip}([1,2], [3,4]) \rightarrow x = [(1,3),(2,4)]$$

This means that the three last lines of code perform a simple dot operation between the two vectors given as argument to `zip()`.

$$\text{It can be rewritten as: } r = \text{np.dot}(u, v)$$

.3

Step 1: find the vector \vec{w}_2 orthogonal to \vec{u}

If we look at Figure 1, we can see that \vec{w}_1 is collinear to \vec{u} , and that \vec{w}_2 is orthogonal to \vec{u} . Moreover, $\vec{v} = \vec{w}_1 + \vec{w}_2$, which means we can easily compute \vec{w}_2 if we have already computed \vec{w}_1 .

$$\vec{w}_2 = \vec{v} - \vec{w}_1$$

Step 2: Make it so the orthogonal vector \vec{w}_2 has the same norm as vector \vec{u}

We must first compute $\|\vec{u}\|$ as well as $\|\vec{w}_2\|$.

By multiplying \vec{w}_2 by a given real value α we can find a new vector \vec{w}_2' that is collinear to \vec{w}_2 , but of different norm.

$$\alpha = \|\vec{u}\| / \|\vec{w}_2\|$$

`def orthogonal_norm_on_first(u, v)` is the function inside `some_script.py` that does this computation.

3 - Computing Eigenvalues, Eigenvectors, and Determinants

.1

$$\text{Det}(A) = \text{Det} \begin{pmatrix} 1 & 1 & 3 \\ 1 & 5 & 1 \\ 3 & 1 & 1 \end{pmatrix}$$

$$1 * \text{Det} \begin{pmatrix} 5 & 1 \\ 1 & 1 \end{pmatrix} - 1 * \text{Det} \begin{pmatrix} 1 & 1 \\ 3 & 1 \end{pmatrix} + 3 * \text{Det} \begin{pmatrix} 1 & 5 \\ 3 & 1 \end{pmatrix} = 1 * (5 - 1) - 1 * (1 - 3) + 3 * (1 - 15) = 4 + 2 - 42 = -36$$

The computer returned the same value.

To compute eigenvalues and eigenvectors we have to compute the characteristic polynomial, $P(A)$:

$$P(A) = \text{Det}(A - I\lambda) = 0 \iff \text{Det} \begin{pmatrix} (1-\lambda) & 1 & 3 \\ 1 & (5-\lambda) & 1 \\ 3 & 1 & (1-\lambda) \end{pmatrix} = 0$$

$$(1-\lambda) * \text{Det} \begin{pmatrix} (5-\lambda) & 1 \\ 1 & (1-\lambda) \end{pmatrix} - 1 * \text{Det} \begin{pmatrix} 1 & 1 \\ 3 & (1-\lambda) \end{pmatrix} + 3 * \text{Det} \begin{pmatrix} 1 & (5-\lambda) \\ 3 & 1 \end{pmatrix} \\ = (1-\lambda) * (4 - 6\lambda + \lambda^2) - 1 * (2 + \lambda) + 3 * (1 - 15 + 3\lambda) = -36 + 7\lambda^2 - \lambda^3 = -(\lambda + 2)(\lambda - 3)(\lambda - 6)$$

This means our eigenvalues are: $\{-2, +3, +6\}$, now we just have to compute the respective eigenvectors.

.2

.3

4 - Computing Projection Onto a Line

.1

.2

We have a function $\alpha : 3x - 2y = -6$ that describes a line, and we have a point $A(5, 4)$. Let's see the steps required to compute the distance between A and f , figure 2 is a sketch of what is explained in the item below:

(1) We can rewrite it as $f(x) = \frac{-6+2y}{3}$

(2) Choose two values for x , and compute the respective images: $f(x_1) = y_1$ and $f(x_2) = y_2$

(3) Compute a vector \vec{u} that is collinear to our function f , $\vec{u} = \begin{bmatrix} x_2 - x_1 \\ y_2 - y_1 \end{bmatrix}$

(4) Compute \vec{v} , vector between a point of the function f and our point A , $\vec{v} = \begin{bmatrix} x_1 - A_x \\ y_1 - A_y \end{bmatrix}$

(5) Compute \vec{w}_1 the projection of \vec{v} onto \vec{u} , \vec{w}_1 and \vec{u} are collinear

(6) Compute $\vec{w}_2 = \vec{v} - \vec{w}_1$, \vec{w}_2 is orthogonal to \vec{u}

(7) Compute $\|\vec{w}_2\|$, which is the distance between the point A and the line represented by the function f

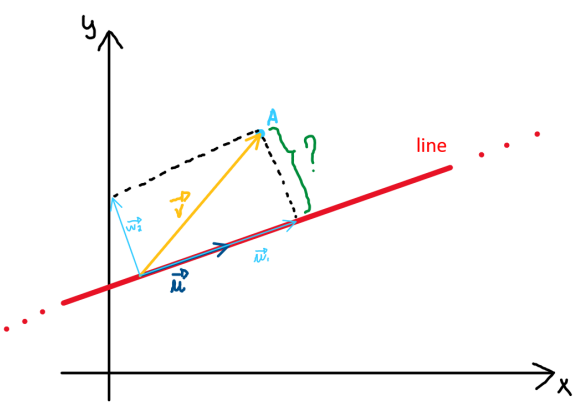


Figure 2: Distance between line and a point A