## An introduction to majorization and some personal thoughts about it

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## Abstract

The following article is *somehow* an introduction to the theory of majorization. Its objective is to present the main ideas of the theory, as well as some personal thoughts about it. The reader can a complete introduction to the theory in [?].

I would assume some basic knowledge in linear algebra, probability theory and familiarity with information theory.

## 1 Introduction

Majorization is a concept that appears in various fields of mathematics whenever we need to talk about non-uniformity, or *randomness*. Many approaches are possible depending on the field, you are working on. A satisfying definition, giving a general intuition of it, is given using the concept of Lorenz curve.

**Definition 1** (Lorenz Curve). Let  $x \in \Delta_n$ , the Lorenz curve  $\mathcal{L}_x$  of x is the plot of the points  $\left(\frac{i}{n}, \sum_{j=1}^{i} x_{(j)}\right)$ , where  $x_{(1)} \geq x_{(2)} \geq \ldots \geq x_{(n)}$  are the components of x in decreasing order.

A Lorenz curve is somehow quantifying the weight of the largest components. In figure 1, the Lorenz curve of the vector (0.5, 0.25, 0.25, 0) is plotted.

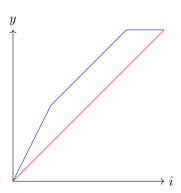
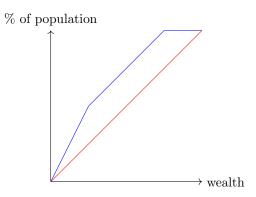


Figure 1: The Lorenz curve of the vector (0.5, 0.25, 0.25, 0.25, 0) in blue and the Lorenz curve of the vector (0.25, 0.25, 0.25, 0.25) in red.



As you can see, the Lorenz curve of the vector x=(0.5,0.25,0.25,0.25,0) is more concave than the Lorenz curve of the vector u=(0.25,0.25,0.25,0.25). How can we interpret informally this fact? According to the figure, the greatest component  $x_{(1)}$  weights more than the greatest component of u. A natural way of representing this is to say that these vectors represent the wealth of a population. x would be a situation where the 25% of the richest people would own the 50% of the wealth, while u would be a situation of perfect equality. In other words, the Lorenz curve of a vector x is one way of representing its non-uniformity. It is quite easy to state that a vector x is less uniform than the uniform distribution but how can we quantify this comparison of non-uniformity? This is where the concept of majorization comes in.

**Definition 2** (Majorization). Let  $x, y \in \mathbb{R}^n$  be two vectors. We say that x is majorized by y, denoted  $x \prec y$ , if  $\mathcal{L}_x$  is below  $\mathcal{L}_y$ , that is,

$$\sum_{i=1}^{k} x_{(i)} \le \sum_{i=1}^{k} y_{(i)} \quad \text{for all } k \in \{1, \dots, n\}$$
 (1)

$$\sum_{i=1}^{n} x_{(i)} = \sum_{i=1}^{n} y_{(i)}.$$
 (2)

**Remark 1.** Here, we have defined majorization for vectors in  $\mathbb{R}^n$  which is not the context of the introduction. However, the definition can be extended and some equivalent definitions can give more intuition of what happens in the general case.

We will, most of the time, work with non-negative vectors. In this case, the introduction is still relevant.

**Example 1.** Let x = (0.5, 0.25, 0.25, 0) and y = (0.25, 0.25, 0.25, 0.25). We have  $\mathcal{L}_x$  below  $\mathcal{L}_y$ , so  $x \prec y$ .

**Example 2.** Consider the set of angles  $\Theta_{\Delta}$  to construct a triangle. Let  $\theta = (\theta_1, \theta_2, \theta_3)$  be a vector of angles. Basic geometry says that  $\theta \in \Theta_{\Delta}$  if and only if  $\theta_1 + \theta_2 + \theta_3 = \pi$ . In other words, if and only if  $\theta \prec (\pi, 0, 0)$ .

This example is quite interesting because it illustrates the comparison of non-uniformity through the comparison of triangles and we will see that this is somehow related to a characterization of majorization.

Here are some basics properties of majorization.

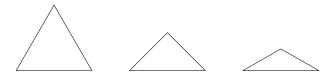


Figure 2: We have respectively  $(\pi/3, \pi/3, \pi/3) \prec (\pi/2, \pi/4, \pi/4) \prec (2\pi/3, \pi/6, \pi/6)$ 

**Proposition 1.** Let  $x, y \in \mathbb{R}^n$ ,  $\prec$  defines a partial order on  $\mathbb{R}^n$ . Moreover,

- If  $x \prec y$  and  $y \prec x$ , then there exists a permutation  $\pi$  such that  $\pi \cdot x = y$ .
- If  $x \prec y$ , then for any permutations  $\pi, \pi'$  we have  $\pi \cdot x \prec \pi' \cdot y$ .
- We have  $|x|u \prec x \prec |x|\delta$  where u is the uniform distribution and  $\delta$  is the Dirac distribution.

Note that  $\prec$  is *indeed* a partial order. There exists vectors x, y such that neither  $x \prec y$  nor  $y \prec x$ .

Before going further, I will present a more information theoretic approach of majorization.

**Definition 3.** A matrix  $A \in \mathbb{R}_+^{n \times n}$  is said to be stochastic if  $u^T A = u^T$ . Moreover, we say that A is doubly stochastic if  $u^T A = u^T$  and Au = u.

A stochastic matrix A can be considered in the finite case as a so-called transition matrix of a Markov chain. These are considered as the most basic representation of channels in information theory. source  $\stackrel{A}{\longrightarrow}$  destination