# Weighted p-majorization and Stable vector

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Abstract - In this paper we generalized the notion of weighed p- majorization for any two vectors in  $\mathbb{R}^n$ . We also define stable vector for a given set of vectors. Then we prove existence of stable vector for n = 2. And finally we give some conditions on a set of vectors for existence of stable vector for  $n \ge 3$ . we also give some examples for illustration our theorems.

Keywords - Majorization, Lorenz Curve, Non-Vanishing

#### INTRODUCTION

The notion of majorization has been introduced while studying the topic such as wealth distribution, inequalities etc. In the early part of the twentieth century, Lorenz introduced a curve, that describes the wealth (or income) distribution of a population. Let  $\alpha = (\alpha_1, \alpha_2, ..., \alpha_n)$  be denote the income profile of a population of size n. Then the line joining origin and the points  $\left(\frac{k}{n},\frac{S_k}{S_n}\right)$ ,  $S_k=\sum_{j=1}^k\alpha_j^{\frac{1}{j}}$ ,  $1\leq k\leq n$ , where  $\alpha_1^{\frac{1}{j}},\alpha_2^{\frac{1}{j}},\cdots,\alpha_n^{\frac{1}{j}}$  be the decreasing rearrangement of the components of \$\alpha\$, is called Lorenz curve of a. If total income or wealth of the population is uniformly distributed among the population then the Lorenz curve is a straight line, otherwise, the curve is convex. Let  $x = (x_1, x_2, \dots, x_n)$ and  $y = (y_1, y_2, \dots, y_n)$  be denote the income of a population of size n, in day 1 and day 2 respectively. If x is majorized by y, then Lorenz curve of x is closer to the straight line of uniform distribution than the Lorenz curve of y. This tech- nique is used in many different [6, 7] such as describing inequality among the size of individuals in ecology, in studies of biodiversity, business modeling, etc. The Lorenz curve also provides a different tool for estimating the distributional dimensions of energy consumption.

Many authors generalized and studied the majorization by introducing different parameters [1, 2, 5]. In 1947, by introducing weighted *p*-majorization for a pair of vectors that are in a similar order (either increasing or decreasing) Fuchs [3], proved an equivalent condition for weighted *p* majorization using continuous convex functions. In 1997, P\*ecar'c and Abramovich [4] discussed an analogs of Fuchs result in which order of one of vector in the pair can be relaxed. In general the wealth (or income) profiles of individuals in a population may

not be in a similar order. This motivates us to define majorization for a pair of vectors in  $\mathbb{R}^n$  by introducing a weight function with out having any order restriction. We use technique of Lorenz as a tool to introduce stability for a given set of points by assigning a proper weight. We prove the existence of a stable vector for a given set of vectors in  $\mathbb{R}^n$  with respect to a given weight. Finally, we give examples to illustrate our techniques.

#### 2. WEIGHTED P-MAJORIZATION

Let  $R_n^+$  be the positive cone of  $R^n$ , the set of all vectors of  $R^n$  with positive coordinates. Let  $p=(p_1,p_2,\cdots,p_n)\in R_n^+$  and  $\alpha=(\alpha_1,\alpha_2,\ldots,\alpha_n)$  be a vector in  $R^n$  with  $\sum_{i=1}^n p_i\,\alpha_i\neq 0$ . Define  $\alpha^p:=(\alpha_1^p,\alpha_2^p,\ldots,\alpha_n^p)$  where  $\alpha_i^p=\frac{p_i\alpha_i}{\sum_{i=1}^n p_i\alpha_i}$   $(1\leq i\leq n)$ . We call such a vector  $\alpha$  as non-vanishing p-vector of  $R^n$ .

**Definition 2.1.** Let  $p = (p_1, p_2, \ldots, p_n)$  be in Rn and  $x = (x_1, x_2, \ldots, x_n)$ ,  $y = (y_1, y_2, \ldots, y_n)$  be two non-vanishing p-vectors of Rn. We say that x is p-weighted majorized by y if  $x^p$  is majorized by  $y^p$ . We denoted it by by  $x <_p y$ .

Let  $p \in Rn$  and x, y be two non-vanishing p-vectors in Rn. If  $x <_p y$ , then from Figure [1] it is to be noted that Lorenz curve of  $x^p$  is closer to the Lorenz curve of uniform distribution than the Lorenz curve of  $y^p$ . In the other words, we say that the vector  $x^p$  is stable than than the vector  $y^p$ .

**Definition 2.2.** Let p be a fixed vector in R n and let  $S = \{x_1, x_2, x_3, \ldots, x_m\}$  denotes a set of any m vectors in R n with non-negative coordinates. A

point  $x_j$  in the set S is said to be a p-stable vector of the set S if  $x_i <_p x_i$  for all  $i \in \{1, 2, ..., m\}$ .

Cumulative percent of income

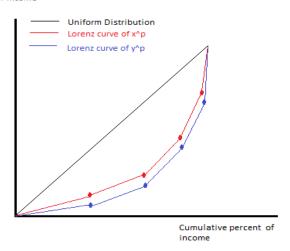


Figure 1: Lorenz curve

**Theorem 2.3.** Let p be a fixed vector in  $\mathbb{R}^2$  and let  $S = \{x_1, x_2, x_3, \ldots, x_m\}$  be a set of any m vectors in  $\mathbb{R}^2$  with positive coordinates. Then there exists a stable vector of the set X with respect to weight p i.e. there exist a vector  $x_i$  o  $(1 \le i_0 \le m)$  such that  $x_i$  or  $(1 \le i_0 \le m)$  such that  $x_i$  o  $(1 \le i_0 \le m)$  such that  $x_i$  o  $(1 \le i_0 \le m)$  such that  $x_i$  or  $(1 \le i_0 \le m)$  such that  $x_i$  or  $(1 \le i_0 \le m)$  such that  $x_i$  or  $(1 \le i_0 \le m)$  such that  $(1 \le i_0 \le m)$  such that

*Proof.* Let  $p = (p_1, p_2)$  and  $x_i = (x_{i1}, x_{i2})$  for  $i \in \{1, 2, ..., m\}$ . Then

$$x_{i}^{p}=(x_{i1},x_{i2})=\left(\frac{p_{1}x_{i1}}{\sum_{j=1}^{2}p_{j}x_{ij}},\frac{p_{2}x_{i2}}{\sum_{j=1}^{2}p_{j}x_{ij}}\right)\,(1\leq i\leq n)$$

Further, for each i  $(1 \le i \le n)$ , there exist a permutation  $\sigma^i$  on  $\{1,2\}$  such that

$$x_{i}^{p\downarrow} = \left(x_{i1}^{p\downarrow}, x_{i2}^{p\downarrow}\right) = \left(\frac{p_{\sigma^{i}(1)}x_{i\sigma^{i}(1)}}{\sum_{j=1}^{2}p_{j}x_{j}^{i}}, \frac{p_{\sigma^{i}(2)}x_{i\sigma^{i}(2)}}{\sum_{j=1}^{2}p_{j}x_{j}^{i}}\right)$$

Consider the set  $\phi = \{\frac{p_{\sigma^i(1)}x_{i\sigma^i(1)}}{\sum_{j=1}^2 p_j x_j^i} \colon 1 \leq i \leq m\}$ . As the set  $\phi$  has a minimum, let it be at  $i_\sigma$  where  $i_\sigma \in \{1,2,\cdots,n\}$ . Thus we get  $x_{io1}^{p\downarrow} \leq x_{i1}^{p\downarrow}$  for all  $i \in \{1,2,\cdots,n\}$ . Therefore  $x_{iO} \prec_\rho x_i$  for all  $i \in \{1,2,\cdots,n\}$ . Hence  $x_{io}$  is p-stable vector of the set S.

The following example shows that for  $n \ge 3$  it is not always possible to find a p-stable vector of a given set of vectors in  $\mathbb{R}^n$  with positive coordinates and a given vector p in  $\mathbb{R}^n_+$ .

**Example** 2.4. Take 
$$n = 3, m = 2, p = (1,1,1), x_1 = \{3,1,1\}$$
 and

 $x_2 = \{2.7, 1.5, 0.8\}$ . Then  $x_1^p = \left\{\frac{3}{5}, \frac{1}{5}, \frac{1}{5}\right\}$  and  $x_2^p = \left\{\frac{2.7}{5}, \frac{1.5}{5}, \frac{0.8}{5}\right\}$ . By a direct calculation, one can show that neither  $x_1^p$  majorized by  $x_2^p$  nor  $x_2^p$  majorized by  $x_1^p$ . Thus neither  $x_1$  or  $x_2$  is a p-stable vector of the set containing the vectors  $x_1$  and  $x_2$  with respect to the weight p.

**Example** 2.5. Take  $n=3, m=2, p=(1,2,3), x_1=\{3,4,5\}$  and  $x_2=\left\{\frac{2,5,14}{3}\right\}$ . Then  $x_1^p=\left\{\frac{3}{26},\frac{9}{26},\frac{15}{26}\right\}$  and  $x_2^p=\left\{\frac{2}{26},\frac{10}{26},\frac{14}{26}\right\}$ . It is easy to verify that neither neither  $x_1^p$  majorized by  $x_2^p$  nor  $x_2^p$  majorized by  $x_1^p$ . Thus neither  $x_1$  or  $x_2$  is a p-stable vector of the set containing the vectors  $x_1$  and  $x_2$  with respect to the weight p.

Let  $p(p_1, p_2, \dots, p_n)$  be a fixed vector in  $\mathbb{R}^n_+$  and  $S = \{x_1, x_2, x_3, \dots, x_m\}$  be a set if vectors in  $\mathbb{R}^n$ . Suppose Suppose  $x_i = (x_{i1}, x_{i2}, \dots, x_{in})$  for  $i = 1, 2, \dots, m$ . Then

$$\begin{array}{l} x_{i}^{p} = \left(x_{i1}^{p}, x_{i2}^{p}, \cdots, x_{in}^{p}\right) = \\ \left(\frac{p_{1}x_{i1}}{\sum_{j=1}^{n} p_{j}x_{j}^{j}}, \frac{p_{2}x_{i2}}{\sum_{j=1}^{n} p_{j}x_{j}^{j}}, \cdots, \frac{p_{n}x_{in}}{\sum_{j=1}^{n} p_{j}x_{j}^{j}}\right), i = 1, 2, \cdots, m. \end{array}$$

For each i, there exist a permutation  $\sigma^i$  on  $\{1,2,\dots,n\}$  such that

$$\begin{split} \boldsymbol{x}_{i}^{\mathfrak{p}\downarrow} &= \left(\boldsymbol{x}_{i1}^{\mathfrak{p}\downarrow}, \boldsymbol{x}_{i2}^{\mathfrak{p}\downarrow}, \cdots, \boldsymbol{x}_{in}^{\mathfrak{p}\downarrow}\right) = \\ &\left(\frac{p_{\sigma^{i}(1)}x_{i\sigma^{i}(1)}}{\sum_{j=1}^{n}p_{j}x_{j}^{i}}, \frac{p_{\sigma^{i}(2)}x_{i\sigma^{i}(2)}}{\sum_{j=1}^{n}p_{j}x_{j}^{i}}, \cdots, \frac{p_{\sigma^{i}(n)}x_{i\sigma^{i}(n)}}{\sum_{j=1}^{n}p_{j}x_{j}^{i}}\right) \end{split}$$

Consider the set  $\phi = \{\frac{p_{\sigma^i(1)}x_{i\sigma^i(1)}}{\sum_{j=1}^2 p_j x_j^i} : 1 \le i \le m\}$ . This set  $\phi$  has a minimum say at  $i_\sigma$  where  $i_\sigma \in \{1,2,\cdots,n\}$ . Thus we get  $x_{io1}^{p\downarrow} \le x_{i1}^{p\downarrow}$  for all  $i \in \{1,2,\cdots,m\}$ . Conditions on vectors....

(1) 
$$x_{i02}^{p\downarrow} \le x_{i2}^{p\downarrow} + \Delta_1^n$$
 for  $i = 1, 2, \cdots, m$ , where  $\Delta_1^i = x_{i1}^{p\downarrow} - x_{i01}^{p\downarrow}$ .

$$\begin{array}{lll} \text{(2)} & x_{i03}^{p\downarrow} \leq x_{i3}^{p\downarrow} + \varDelta_2^n & \text{for} & i = 1, 2, \cdots, m, & \text{where} \\ \varDelta_2^i = \sum_{j=1}^2 x_{ij}^{p\downarrow} - \sum_{j=1}^2 x_{i0j}^{p\downarrow}. & \end{array}$$

(3) 
$$x_{i04}^{p\downarrow} \le x_{i4}^{p\downarrow} + \Delta_3^n$$
 for  $i = 1, 2, \cdots, m$ , where  $\Delta_2^i = \sum_{j=1}^3 x_{ij}^{p\downarrow} - \sum_{j=1}^3 x_{i0j}^{p\downarrow}$ .

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$$(n-2)x_{i0(n-1)}^{p\downarrow} \leq x_{i(n-1)}^{p\downarrow} + \varDelta_{n-2}^n \quad \text{ for } \quad i=1,2,\cdots,m,$$

where 
$$\Delta_{n-2}^i = \sum_{j=1}^{n-2} x_{ij}^{p\downarrow} - \sum_{j=1}^{n-2} x_{i0j}^{p\downarrow}$$
.

If the vectors satisfy these conditions then we get a stable vector say  $x_{io}$  with respect to p.

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